

**EVALUATING CPU UTILIZATION IN A CLOUD
ENVIRONMENT**

BY
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MASTER OF SCIENCE

In

COMPUTER SCIENCE

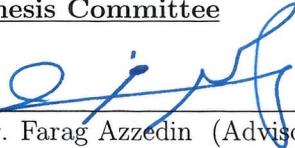
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This thesis, written by **HAFIZ UR RAHMAN** under the direction of his thesis adviser and approved by his thesis committee, has been presented to and accepted by the Dean of Graduate Studies, in partial fulfillment of the requirements for the degree of **MASTER OF SCIENCE IN COMPUTER SCIENCE**.

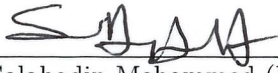
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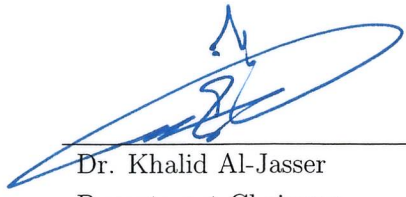
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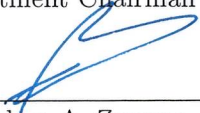
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Dedication

To my parents and family, may Allah protect them and give them good health and happiness.

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LIST OF ABBREVIATIONS

CMP	Chip Multiprocessor
CPU	Central Processing Unit
FV	Full Virtualization
KVM	Kernel-based Virtual Machine
HAV	Hardware Assisted Virtualization
IaaS	Infrastructure-as-a-Service
NUMA	Non-Uniform Memory Access
OS	Operating System
PaaS	Platform-as-a-Service
PV	Paravirtualization
pCPU	Physical CPU
PTS	Phronix Testing Suite
RHEV	Red Hat Enterprise Virtualization
SaaS	Software-as-a-Service
SMP	Symmetric Multiprocessing
vCPU – VM	Virtual CPU to Virtual Machine
vCPU – pCPU	Virtual CPU to Physical CPU
VM	Virtual Machine
VMM	Virtual Machine Monitor

ABSTRACT

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Virtualization is used to ease computing resource management, resource utilization and running multiple heterogeneous or homogeneous operating systems on a single physical machine. Virtualization offers many advantages such as reducing fiscal costs, easing system migration, running legacy applications, easing backups and disaster recovery, and utilizing the shared resources over traditional data centers. In virtualized environments, the higher the percentage of the CPU utilization results in the maximum performance. However, virtualization systems have shortcomings including virtual machines allocation, virtual CPU configuration, and virtual CPU to physical CPU mapping. Such shortcomings may lead to system performance degradation. In this thesis, we focus on the role that hypervisors, virtual CPU to virtual machine allocation, and the virtual CPU to physical CPU mapping play on CPU utilization. PTS benchmarking tool is used as a traffic generator and analyzer, which provides a detailed evaluation of CPU utilization. The results of this study will help cloud service providers and researchers select and decide which virtualization technology, virtual CPU to virtual machine configuration, and mapping strategies to used for better performance and best CPU resources allocation.

ملخص الرسالة

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الافتراضية مستخدمة لتسهيل إدارة موارد الحوسبة ولتسهيل الاستخدام الأمثل لتلك الموارد ولتشغيل عدة أنظمة تشغيل متجانسة أو متفاوتة على ماكينات مادية وحيدة. الافتراضية تقدم العديد من المزايا مثل تقليص التكاليف المالية وتسهيل النسخ الاحتياطية والتعافي من الكارثة والاستخدام الأمثل للموارد المشتركة من مراكز البيانات التقليدية. في البيئات الافتراضية، كلما زادت نسبة الاستفادة من وحدة المعالجة المركزية (CPU) فإنها تسفر عن الأداء عالي لها. ومع ذلك، الأنظمة الافتراضية لديها قصور مشتملة على تخصيص الآلات الافتراضية، تكوين وتهيئة وحدة المعالجة المركزية الافتراضية، ووصل وحدة المعالجة المركزية الافتراضية بالمادية. هكذا قصور ربما يقود إلى تراجع في أداء النظام. في هذه الرسالة، سنركز على الدور للـ Hypervisor بالإضافة إلى وصل وحدة المعالجة المركزية الافتراضية - بوحدة المعالجة المركزية المادية في أن تلعب دوراً في الاستخدام الأمثل لوحدة المعالجة المركزية. أداة الاختبار (PTS) مستخدمة كمولد ومحل لحركة المرور، والتي تزود بتقييم مفصل لتكشف التباين في الاستخدام الأمثل لوحدة المعالجة المركزية. نتائج هذه الدراسة ستساعد مزودي الخدمة السحابية والباحثين ليختاروا ويقرروا التقنية الافتراضية الأفضل، التهيئة والتكوين لوحدة المعالجة المركزية الافتراضية في الآلات الافتراضية، وربط الاستراتيجيات لاحتياجاتهم الخاصة.

CHAPTER 1

INTRODUCTION

Virtualization is the fundamental element of cloud computing [6, 7, 13] by which we can deliver resources or data as a service. In the last decade, virtualization has attracted many different research groups working on server consolidation, security, and computing [8, 9]. For example, distributed data centers are now being utilized by using virtualization technology which was not possible in the past.

Usually, servers are underutilized, which results in making the data centers unproductive [10]. Traditional data centers have several shortcomings including maintenance cost and resources utilization. Furthermore, traditional data centers suffer from server proliferation, low resource utilization, physical infrastructure cost, and migration challenges [3, 8, 11, 13]. Thus, virtualization plays a vital role in mitigating such challenges [8]. Virtualization makes use of server resources in a well-organized manner by setting up different servers within different cloud types [13]. Moreover, virtualization consolidates workloads so that one physical machine can be used for different users. This improves the efficiency and utilization of

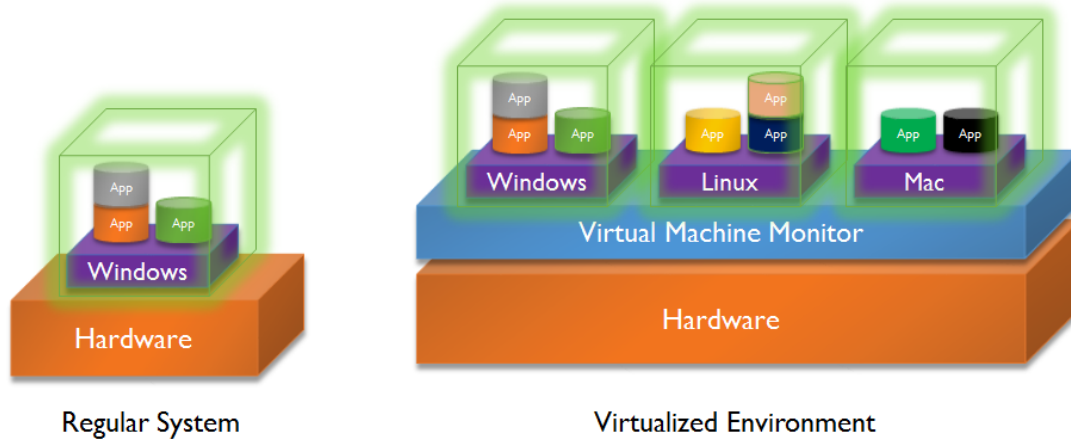


Figure 1.1: Regular system and Virtualized environment

data centers by allowing more work to be done on a smaller set of physical nodes [14, 15]. Through virtualization, organizations can achieve better management and improve resource efficiency [15]. As a result, organizations can access and manage their data more efficiently. Therefore, many organizations are adopting virtualization technology to reduce the cost while maximizing the productivity, flexibility, responsiveness, and efficiency.

Virtualization can be done to various resources such as CPU, memory, or I/O devices. Virtualization vendors use different technologies to provide virtualization environment. These environments can be built to enhance the utilization of traditional IT infrastructures as well as resource management capabilities. Hypervisors are used in virtualized environments as agents facilitating virtual machines and hardware [4, 15, 16, 17].

1.1 Problem Statement

Virtualization technology inserts an additional abstraction layer between the hardware and the operating system [4, 8, 18] as shown in Figure 1.1. In a regular system, the hardware resources are shared by single OS. While in virtualization environments, hypervisors are responsible to manage hardware resources and virtual machines. Moreover, every guest operating system is in charge of virtual resources and concurrently share and access the hardware resources [9]. Therefore, virtualization systems face challenges such as virtual machines (VMs) allocation, virtual CPU allocation, and virtual CPU to physical CPU (vCPU-pCPU) mapping. Such challenges may lead to system performance degradation [19, 20, 21].

Nevertheless, CPU is one of the most significant and critical resource among all the available resources in a system [8, 22, 23]. Improving CPU utilization is one goal of any virtualization technology. In addition, CPU utilization is used as a metric to measure system performance. The cloud provider assigns resources to each user request aiming to minimize resource allocation and fulfill user requirements [24]. The following [25] are the main factors that have implications on CPU utilization namely: resource sharing between physical cores, resource sharing between logical cores (hyper threading), the choice of guest OS, the BIOS setting for power management (machine level), the hypervisor/OS setting for power management, and VM setting (over-commitment between vCPU and physical CPU (pCPU)).

CPU utilization can be improved at different levels by tuning up different

components. At VM level [19, 20, 21, 26], the following parameters can be tuned to improve the CPU utilization: (a) vCPU-pCPU commitment, (b) the OS setting for power management, and (c) the OS scheduling algorithm. Moreover, at hypervisor level, the following parameters can be tuned to improve the CPU utilization: (a) the hypervisor scheduler and (b) vCPU-pCPU mapping [19, 21]. Furthermore, configuring BIOS setting for power management [25] and choosing the state-of-the-art hardware, such as Intel and AMD virtualization aware processors [4, 8, 18] that provides the necessary support for virtualization, contribute towards improving CPU utilization.

In this thesis, we evaluate the effects of hypervisor, virtual machine allocation, vCPU-VM configuration, and vCPU-pCPU mapping on performance in terms of CPU utilization in a cloud environment. Various experiments for CPU utilization are performed using Citrix xenServer (commercial) and KVM (open source) hypervisors. Phronix Testing Suite (PTS) [27] benchmarking tool is used as a traffic generator and analyzer. We chose PTS, since it tracks CPU performance and generates data points for performance investigation. The aim of these experiments is to investigate how to allocate CPU resources (in terms of cores and logical CPUs) to achieve better performance.

1.2 Motivation

In this thesis, we were motivated by the fact that virtualization suffers from drawbacks. In addition, researchers evaluate and analyze hypervisors without inves-

Investigating the best vCPU-VM configuration through which better CPU utilization and performance for each hypervisor can be expected. They assigned vCPU to VM based on non-suitable vCPU-VM configuration [15, 28, 29, 30]. Furthermore, there are a variety of vendors for the virtualization environment and all of them claim that their virtualization hypervisor is the best, however they depend on the used application [17]. Moreover, when applying virtualization technology to an infrastructure environment, which hypervisor among others is better and faster in terms of CPU utilization? When a system administrator or a researcher deploy virtual machine in a cloud environment, which vCPU-VM configuration and mapping strategies for virtual machine is the best to improve CPU utilization? What is the effect of VM and vCPU on performance? Is there any other setting or configuration by which CPU utilization be improved?

1.3 Objectives

In this research study, our aim is to investigate the best CPU resources allocation and better performance using the latest version of the state-of-the-art hypervisors. In addition, we investigate the effect of hypervisor, VM, vCPU, and pinning strategies on performance. We also present the best vCPU configuration and pinning strategy for VM in a cloud environment through which better performance and maximum CPU utilization can be expected.

Our finding will be a road map to assist cloud service providers to choose the best hypervisor, vCPU-VM configuration, and mapping strategies for their

specific needs.

In a nutshell, the thesis objectives are:

1. Build a private cloud using the latest version of commercial and open source hypervisors (Citrix xenServer and KVM).
2. Prior to evaluate the hypervisors, investigate the best vCPU - VM configuration and vCPU pCPU mapping strategies, through which better CPU utilization can be expected for each hypervisor.
3. Evaluate the effects of Hypervisor, Virtual Machine, vCPU VM assignment, and vCPU pCPU mapping on performance in terms of CPU utilization.

1.4 Contributions

Based on the objectives defined previously, we outline the main contributions of this thesis as follows:

Building a private cloud using the latest version of commercial and open source hypervisors (Citrix xenServer version 7.0.1 and KVM version 4.4.0). In this research work, we focus on the effect virtualization technology, VM allocation, vCPU-VM configuration, and vCPU-pCPU mapping to find out the variation in CPU utilization prior to evaluate the hypervisors.

As another contribution, we provide recommendations for vCPU-VM configuration and vCPU-pCPU mapping through which better CPU utilization on each hypervisor can be achieved. As a result, cloud service providers will get the ben-

efits, when they deploy VMs in a cloud environment or evaluate open source and commercial hypervisors (KVM, Citrix xenServer, VMware, and Hyper-V).

1.5 Thesis Organization

This thesis is organized into six chapters. Chapter 2 presents a background on virtualization technology and outlines the definition of virtualization terminology. Chapter 3 provides an extensive literature review of the impact of hypervisor, vCPUs-VMs configuration, and vCPU-pCPU mapping on performance. Chapter 4 describes our research methodology and experimental setup. We analyze the evaluation results in chapter 5 and draw conclusions in chapter 6. Furthermore, we summarize our major contributions, discuss perspectives, and envisions future work directions in chapter 6 as well.

CHAPTER 2

BACKGROUND

This chapter introduces the concept of cloud computing and virtualization. For completeness and clarity purposes, we provide a list of definitions that will be used throughout this thesis. We also outline the main concept of virtualization and cloud computing. Finally, we highlight the virtualization approaches and challenges.

1. **Cloud Computing:** Cloud computing is a pay-per-use model on demand delivery of computing resources over the internet [31]. Figure 2.1, shows the cloud computing environment.
2. **Virtualization:** Virtualization refers to a variety of different computing concepts. It is used for ease computing resource management, resource utilization, and running multiple heterogeneous or homogeneous OSs on a single physical machine [7, 9].
3. **Hypervisor:** An abstraction layer of software which is also known as Vir-

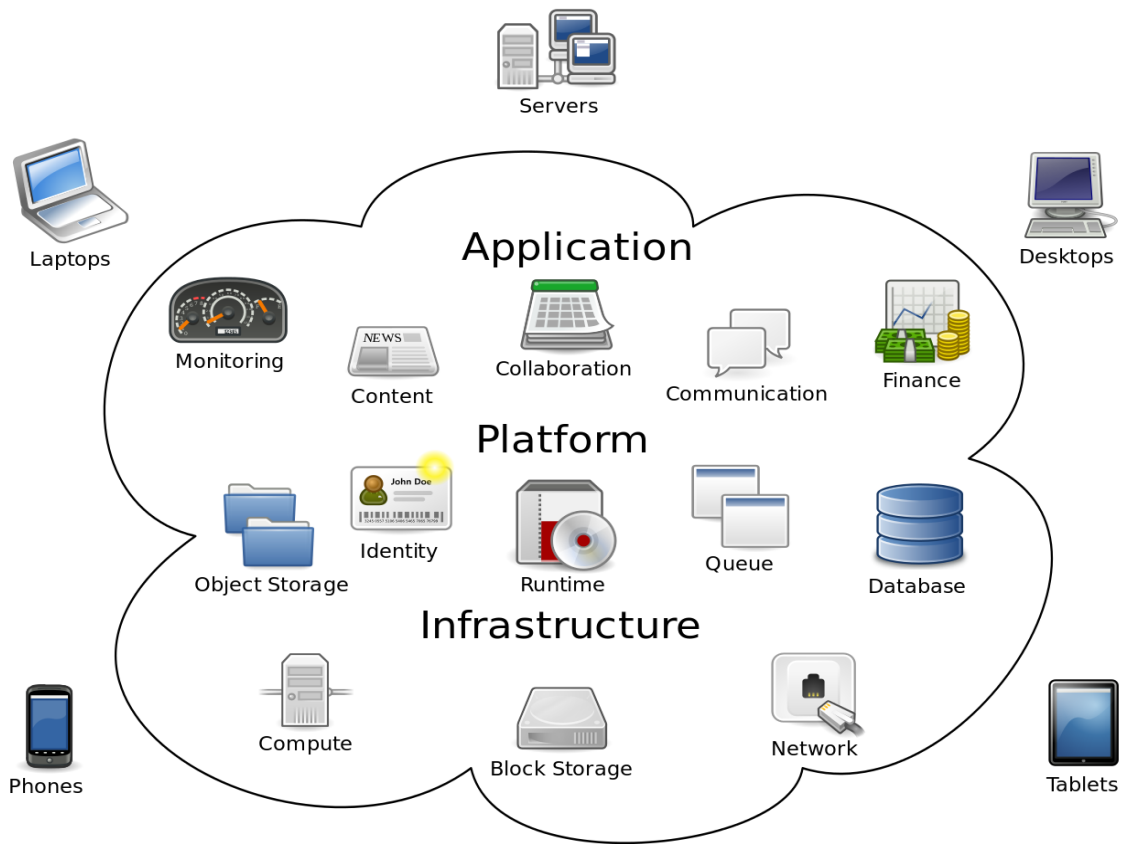


Figure 2.1: Cloud Computing [1]

tual Machine Monitor (VMM) that makes virtualization possible. A widely accepted definition has been provided by [8] for VMM: “As a piece of software a VMM has three essential characteristics. First, the VMM provides an environment for programs which is identical to the original machine; second, programs running in this environment show at worst only minor decreases in speed; lastly, VMM is a complete control of system resources”.

4. **Virtual Machine:** A Virtual Machine (VM) is an efficient isolated duplicate of the real physical machine that can has its own OS [22, 32].

5. **Host Machine:** A host machine is a server component or physical machine on which a hypervisor is running. Host machines provide the computing resources such as processing power, memory, network, I/O, etc., to support virtual machines [8].
6. **Guest OS:** A guest OS is an independent instance of an OS that is installed and run in a virtual machine.
7. **Physical CPU:** A physical CPU (pCPU) is the central processing unit responsible of all processing operations. A server can have more than one processor.
8. **Core:** Inside physical CPU, there are more than one operations unit (like a processor), called core [33].
9. **Logical CPU:** It is a processing unit that is capable of executing its own thread in parallel with other logical cores. Number of logical processors means number of threads. Normally, a core can handle one thread at the same time (processor time slot). But when Hyper-Threading is activated and supported, the core can handle two threads at the same time [8, 22].
10. **Virtual CPU:** A virtual CPU (vCPU) is also known as a virtual processor. A vCPU is a physical central processing unit that is assigned to a virtual machine. A vCPU represents a portion or share of a physical CPU that is assigned to a virtual machine [8, 22].
11. **CPU Utilization:** CPU utilization refers to a computer's usage of pro-

cessing resources, or the amount of work handled by a CPU in a given time. Actual CPU utilization varies depending on the amount and type of managed computing tasks [34]. It is the metric that represents how busy is a processor. It shows whether the CPU on the host is completely occupied. Moreover, It can be used to track CPU performance regressions or improvements.

2.1 Cloud Service Models

The cloud computing resources can be accessed through a set of service models. The cloud service models describe how cloud services are made available to clients [35]. Infrastructure-as-a-Service (IaaS), Platform-as-a-Service (PaaS), and Software-as-a-Service (SaaS) are examples of cloud computing service models [3, 8]. These service models can be selected and customized for an organization's requirements [31, 36]. Cloud service models are shown in Figure 2.2 and details are given below for each service model.

2.1.1 Infrastructure-as-a-Service

IaaS is the most straightforward model for delivering cloud services. This model provides the power to control and manage the fundamental computing resources such as virtual machines, physical servers, operating system, storage, network, firewalls, and so on. By using IaaS service, users are able to develop their own application environments. Furthermore, IaaS allows an organization to rent com-

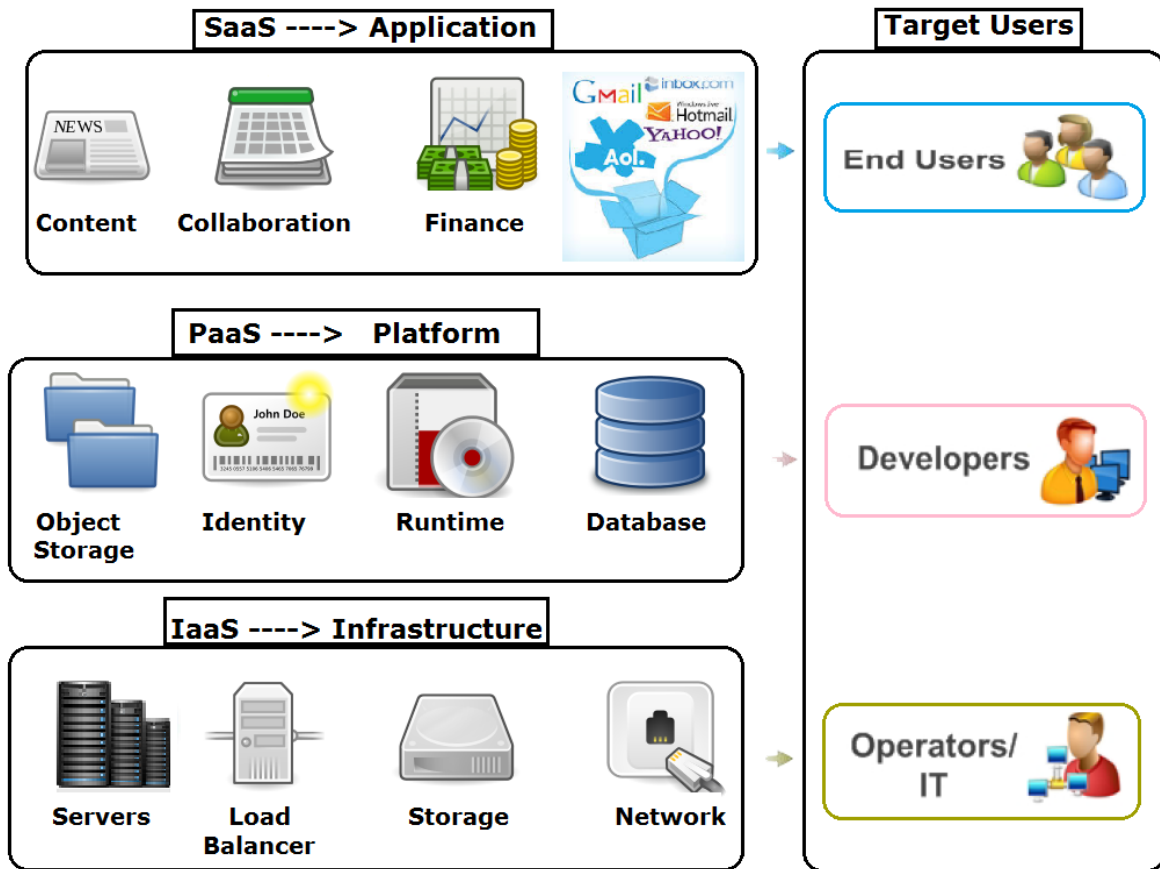


Figure 2.2: Cloud Service Models

puting resources on demand instead of spending much money on management and building their own data centers [3, 8, 28, 35]. Amazon EC2, Nimbus, and Eucalyptus are popular IaaS providers [28].

2.1.2 Platform-as-a-Service

PaaS hides the low-level details from the user and provides a pre-built application development platform for the client to develop, deploy, and manage cloud applications while not worrying about the technology. PaaS allows a user to deploy applications but hides from the user the backend infrastructure such as OS

[3, 8, 35]. Google AppEngine, Heroku, and Microsoft Azure are the most popular PaaS providers [31].

2.1.3 Software-as-a-Service

SaaS is the top level of service models which providing software solutions to users. A user can access the SaaS applications using any thin interface such as mobile device or browsers. Google documents, google apps, online mail, project managements systems, and social media platforms are some examples of SaaS [3, 8, 35].

The main difference between SaaS and PaaS is that SaaS delivers already complete and developed applications while PaaS represents a platform where the user has a complete control over the applications [3, 8].

2.2 Cloud Deployment Models

The cloud service model can be deployed on one or more deployment models. There are four different approaches to deploying cloud service models, namely private cloud, public cloud, hybrid cloud, and community cloud [3, 8] as shown in Figure 2.3. Each of these deployment models is explained in the next subsections.

2.2.1 Private Cloud

A private cloud refers to cloud infrastructures in which the services are maintained on a private network and not available for public use. A private cloud is hosted in the data center of a company and provides its services only to users inside

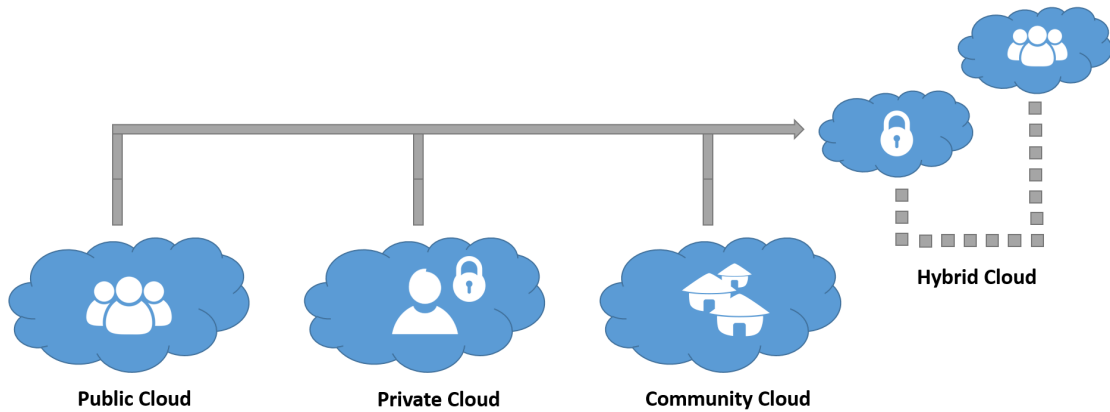


Figure 2.3: Cloud Deployment Models [2]

that company or its partners. The cloud infrastructure is accessed only by the members of the organization and/or by granted third parties. A private cloud has the potential to give the organization greater control over the infrastructure and computational resources [37, 38]. Private clouds can be managed by the organization itself or by a third party cloud [31, 36].

2.2.2 Public Cloud

Public clouds refer to infrastructures operated and owned by a cloud service provider such as Microsoft Azure, IBM’s Blue Cloud, and Google AppEngine are the examples of public clouds [31, 36]. Public clouds provide services that are open to users over the Internet, for a pay-per-usage fee. In public clouds, users can scale their use on demand and do not need to purchase hardware to use the service [37, 38].

2.2.3 Hybrid Cloud

Hybrid clouds are more complex than the other deployment models, since they involve a composition of two or more clouds (private, community, or public). A hybrid cloud is a composition of at least one private cloud and at least one public cloud. A hybrid cloud is typically offered in one of two ways: a vendor has a private cloud and forms a partnership with a public cloud provider, or a public cloud provider forms a partnership with a vendor that provides private cloud platforms [38]. This deployment model gives greater flexibility, scalability, and security i.e., by allowing workloads to move between private and public clouds [31, 36].

2.2.4 Community Cloud

A community cloud is shared by multiple organizations to support a specific community with shared concerns i.e., security, compliance, jurisdiction, etc. [31, 36]. A community cloud falls between public and private clouds with respect to the target set of consumers. It is somewhat similar to a private cloud, but the infrastructure and computational resources are exclusive to two or more organizations that have common privacy, security, and regulatory considerations, rather than a single organization [38].

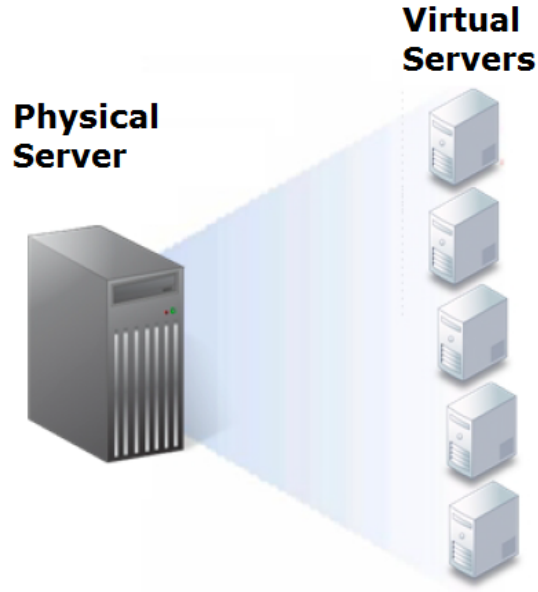


Figure 2.4: Server Virtualization [3]

2.3 Virtualization Solutions

Virtualization is the fundamental element of cloud computing [9, 7] by which we can deliver resources or data as a service. It refers to the creation of virtual resources on a physical one such as server, OS, network, and storage. Different resources can be virtualized including a server, file, storage, and a network virtualization. All of these cases share the same idea, that is physical resources abstraction.

Server virtualization is the most common resource abstraction technique in cloud computing. Server virtualization is a technique in which a physical server can be split into a number of virtual nodes and run multiple concurrent operating systems on the same server to better consume, utilize, and achieve maximum performance [9, 10, 39]. Figure 2.4 represents server virtualization. This kind of

virtualization allows multiple isolated virtual servers to run on a single physical server and can be implemented in different ways. There are three main implementation approaches for server virtualization called full virtualization (FV), paravirtualization (PV), and hardware assistance virtualization (HAV) [3, 4, 8, 35] as shown in Figure 2.7

2.3.1 Hypervisor

A hypervisor, as shown in Figure 2.5 is the core component of a virtualization solution [8, 30]. The hypervisor is responsible for running multiple virtual machines on a host machine to consolidate the workloads so that one physical machine can be multiplexed for many different users. This improves the efficiency of the overall system. The resources such as memory, OS, CPU, network, and data are shared among different virtual machines and completely managed by the hypervisor [40].

This highlights three basic properties or essential requirements for VMM [30]. The first property of the VMM is *equivalence* by which a program running under VMM should have the same behavior and effect as that of the program running on the original physical host machine. The second property is *efficiency* meaning that the virtual CPU instruction is executing directly on host hardware without any intervention by the hypervisor. *Resource Control*, being the last property of VMM. This property enables the VMM to control the allocated resources in such a way that an arbitrary program: (a) can not affect system resources and (b) it does not have access to any resources not explicitly allocated to it. Figure

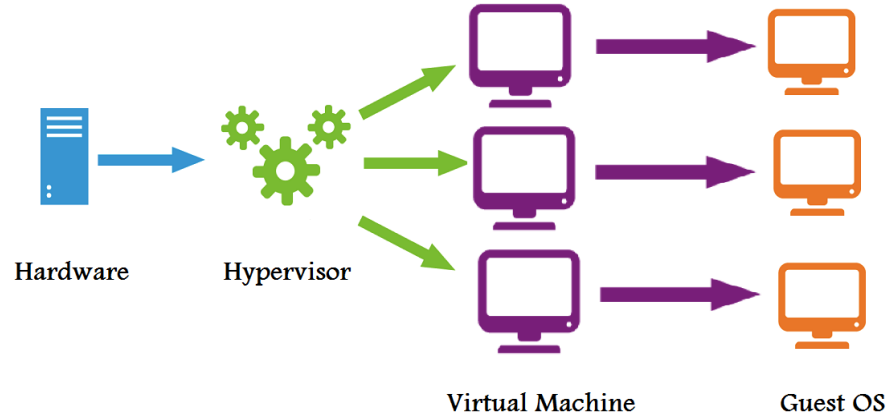


Figure 2.5: A Virtualization Solution

2.5, shows the structure of a simple hypervisor with other essential components.

2.3.2 Types of Hypervisors

Hypervisor are classified into two types [3, 8, 40]; Type I referred to as ***Bare Metal*** and Type II known as ***Hosted***. Figure 2.6, shows the structure of both hypervisors.

Type I

Type I hypervisor is considered a native or bare-metal installation, which means that the hypervisor is the first entity installed on the host machine. Type I hypervisor runs directly on top of the host machine to control access to the hardware and is located right below virtual machines to manage the guest OSs [3, 8, 40].

The most important characteristic of Type I hypervisor is performance. Because there is no intermediary layer between the hypervisor and the physical hardware. Since Type I hypervisors are written specifically to support virtualiza-

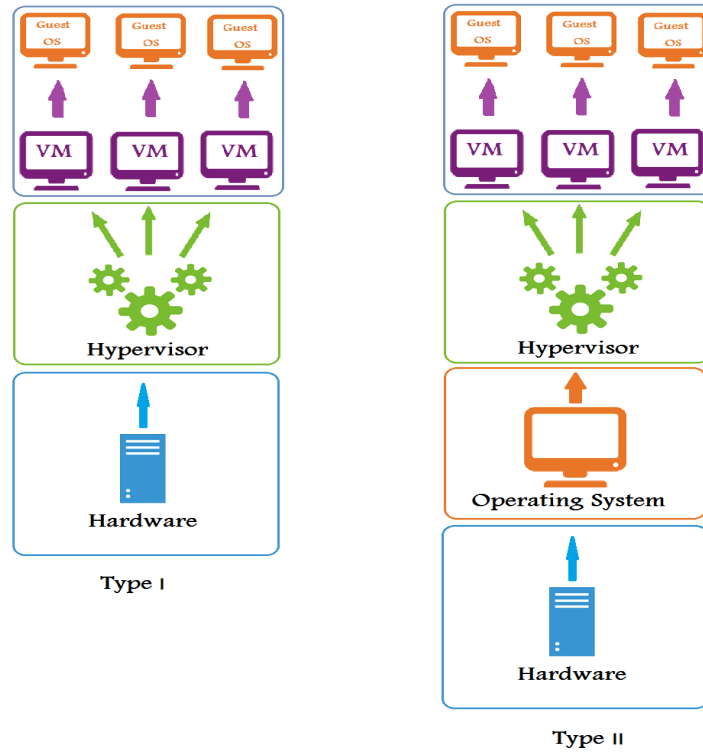


Figure 2.6: Type I and Type II Hypervisors

tion, they usually have a very small footprint compared to general purpose OSs. At the same time, designed specifically to support only virtualization. Type I hypervisors enable us to provide most of the physical hardware resources of the host machine to the guest virtual machines. In general, Type I hypervisors such as VMware ESXi are less compatible with hardware equipment due to their small footprint.

Taking into consideration the performance and capabilities built into Type I hypervisors, they are good candidates to run in data centers [3, 8, 40]. These hypervisors are high performance with special capabilities (e.g high availability, dynamic resource management) designed to support guest OSs that require high availability features. This makes Type I hypervisors a more efficient solution with

better performance compared to Type II hypervisors. VMware vSphere ESXi, Microsoft Hyper-V, Citrix xenServer, Red Hat Enterprise Virtualization (RHEV), and KVM are some examples of Type I hypervisors.

Type II

Type II hypervisors, also known as hosted hypervisors. They are not deployed in bare-metal fashion [3, 8, 40]. They are installed as an application on top of a traditional OS. This type was the first x86 hypervisor to leverage the pre-existing OS installed on the hardware for managing hardware resources. So, it was the fastest way to introduce virtualization into the market.

Type II hypervisors are installed in the form of an application on commodity OSs such as Microsoft Windows and Linux. Furthermore, they are more compatible with physical hardware devices compared to Type I hypervisors. This is because this kind of hypervisors utilize the hardware. Moreover, a Type II hypervisor runs as an application on top of the OS, there is an extra layer between the hypervisor and the hardware.

The difference with the Type I hypervisor is that since a Type II hypervisor does not have direct access to hardware [30], it has to go through one additional cycle and provide the request to the OS. So, every transaction in a Type II hypervisor requires two additional steps, which require more time and processing overhead compared to a Type I hypervisor. Since Type II hypervisors run atop the OS, any issues such as malware, OS failure, and bad device driver will be noticed by all virtual machines running on top of the OS.

Due to the performance and limitations of Type II hypervisors [3, 8, 40], they are not used in data centers. At the same time, Type II hypervisors can be used by application developers, who need to access a number of different OSs on their local machines for research or software testing in different environments [8]. VMware Workstation, Microsoft Virtual PC, Parallels Workstation, and Oracle Virtual-Box are some example of Type II hypervisors provided by leading virtualization vendors [3, 40].

2.4 Virtualization Approaches and Challenges

The x86 platform is the dominant and most commonly used architecture in data centers [3, 8]. They offer different levels of privilege levels (i.e., 0, 1, 2, and 3) known as rings [4, 8, 18], numbered 0 (most privileged) to 3 (least privileged), and main resources such as to execute CPU instruction, memory, and I/O ports are being protected. The ring 0 have the highest privilege with complete control over the system where the OSs kernel is running and control the system. Moreover, they consider themselves the owner of the system resources. Ring 3 is the outermost ring with the most limited level privilege is where the user applications execute and called Userspace or Userland. While the other two rings are rarely used where device drivers reside. This architecture ensures ring 3 application can not make privilege system calls [18]. If they access privilege resources they create an exception and corresponding catch, as a result of the unprivileged state change to privilege mode and the OS execute the instruction.

In virtualization environment Type I hypervisors run in Ring 0 and control the overall system resources like OS. The VM behaviors is exactly the same as the user application program [4, 8, 18]. In virtualized environments, thy system resources including CPU, main memory, and I/O are virtualized to the guest OS. However, the x86 platform was not designed with virtualization support. It was designed to execute a single instance of OS. By running a hypervisor in the most privilege level without modifying the guest OS is a big challenge in virtualization. Because the hypervisor moves up the OS to a ring above and de-privilege the OS. Moreover, the hypervisor can not allow the guest OS to run at the level from which it originally designs because doing so might corrupt the hypervisor code and data [4, 29].

In order to handle the challenges of x86 platform virtualization, there are three main alternative approaches: Paravirtualization (PV) or OS assisted virtualization, Full Virtualization (FV), and Hardware Assisted Virtualization (HAV) [4, 8, 18, 29]. Each of the solutions is shown in Figure 2.7. We discuss these three alternative virtualization approaches in the following subsections.

2.4.1 Virtualization Techniques

In order to handle the challenges of x86 platform virtualization, we are presenting the three commonly used approaches namely FT, PV, and HAV. Figure 2.7, depicts the different approaches used to provide the virtualization layer [4].

Full Virtualization

The hypervisor runs in Ring 0 with binary translation of privileged instructions with the guest OS is running in Ring 1 in unmodified form or in isolation in full virtualization [4, 8, 18] as shown in Figure 2.7. In addition, the guest OS is unaware that it is running in a virtualized environment. Moreover, it did not require OS or hardware assistant to virtualize the underlying hardware or to run the confidential instructions. Because full virtualization used binary translation method in which it dynamically translate the nonvirtualizable instructions to codes, then the translated codes are executed in Ring 0 with the same sequence of the instructions having the same effect on the underlining virtualized hardware.

Due to a full control of the hardware resources and no modification on the part of guest OS, full virtualization enables the hypervisor to support a wide range of OSs and provides the best security and seamless migration and portability of the guest OS. Furthermore, a high performance can be archive by using the binary translation with of the direct execution of OS instruction [4, 8, 18].

Paravirtualization

In paravirtualization (PV) as shown in Figure 2.7, the hypervisor runs directly on the top of hardware as an abstraction layer called virtualization layer, while the guest OS runs within its expected ring 0 for which it originally designed [4, 8, 18]. Furthermore, PV provides hypercall interface for nonvirtualizable instructions to handle and replace kernel operation. But in PV, the guest OS kernel code is

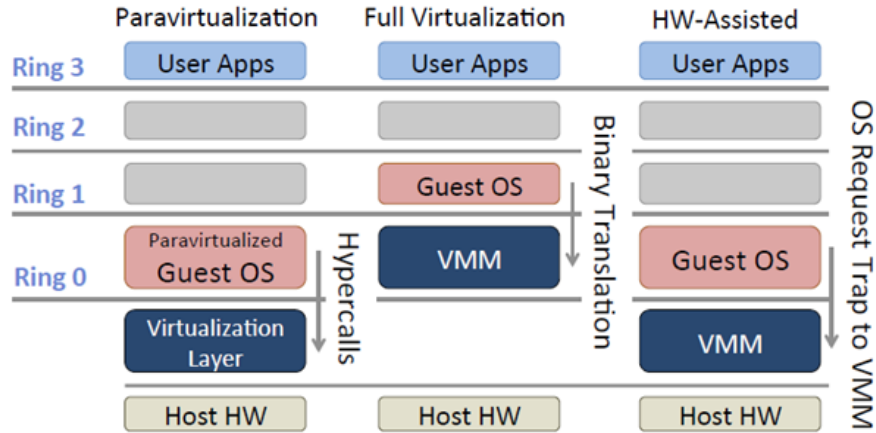


Figure 2.7: Commonly used Approaches for Virtualization Layer [4]

modified and it directly communicates with the hypervisor with hypercalls for better performance and efficiency. In PV, the guest OS is aware of virtualization.

Through kernel code modification, virtualization overhead can be minimized and we can obtain better performance as compared to full virtualization binary translation. However, it is possible to modify open source OS kernel code such as Linux, but it is hard and impossible to modify OS kernel code of other OSs such as Microsoft. Eventually, paravirtualization does not support a wide range of OSs. In addition, PV can not modify older version of open sources OS kernel codes that are already in use.

Hardware Assisted Virtualization

Hardware Assisted Virtualization (HAV) is the latest virtualization technology. In HAV, the guest OS runs in isolation without modifying its kernel code. Latest processors provide the necessary support for virtualization in HAV [4, 8, 18]. Since 2006, hardware vendors (Intel and AMD) produce virtualization aware processors

and provided a new CPU execution mode known as root mode [41]. In root mode they hypervisors run directly on the top of hardware to take control of both hardware and guest OS in a new layer below ring 0 as depicted in Figure 2.7. Furthermore, OS sensitive and privileged requests are automatically trapped by the hypervisor and binary translation, hypercall are no longer required [4, 8, 18].

Intel and AMD produce virtualization aware processors that simplify virtualization. Both manufacturers provide a new CPU execution mode known as root mode that enabled the hypervisor to run in a layer below ring 0 to take control of the guest OS. Using HAV, sensitive and privileged instructions are automatically trapped by the hypervisor removing the need for binary translation or paravirtualization [18]. In addition, HAV provide excellent compatibility to support a wide range of OSs regardless of kernel modification. HAV, can not be used on old x86 platform and requires second generation hardware (virtualization assisted hardware). Citrix xenServer, Hyper-V, KVM, and VMware ESXi hypervisors support hardware assisted virtualization.

CHAPTER 3

RELATED WORK

In the next two subsections, we discuss recently published papers in order to investigate the impact of vCPU-pCPU matching, vCPU-VM assignment, and the impact of hypervisors on CPU utilization. Furthermore, we study the implementation environments used to conduct the performance evaluation. Other factors such as the nature of virtualization type (i.e., full virtualization, para virtualization, or hardware assistance virtualization) are also investigated and summarized.

3.1 Hypervisor Evaluation

Charles David [28] analyzed two types of virtualization namely paravirtualization and hardware assisted virtualization using open sources virtualization platforms KVM (RHEL 5.3 64bit) and Xen 3.1.2 hypervisors on CMP system. The author measured the throughput and overall performance of the hypervisors using PTS benchmarking tool under various levels of workload and compared different system attributes including CPU usage, memory access rate, and I/O operations.

Unfortunately, the author randomly assigned vCPUs to VMs which caused performance degradation. The author did not analyze the root cause of the performance degradation.

Babu et al. [29] evaluated the system performance of three hypervisors. The authors had opted Xen-PV, OpenVZ, and XenServer for para virtualization, container virtualization, and full virtualization, respectively. They compared the performance of these techniques based on Unixbench benchmarking tool. They observed that the hypervisor which supports full virtualization has a comparatively higher system performance in terms of file copy, pipe based context switching, process creation, shell scripts, and floating point operation than the other two virtualization types. However, the authors did not investigate the effects of vCPU-VM and vCPU-pCPU. Moreover, the authors only used one virtual machine for their evaluation.

C. Mancas [30], used Passmask benchmarking tool and evaluated VMware and KVM hypervisors for CPU, memory, and I/O performance. They observed that overall VMware behaves better than KVM. However, there are cases, such as memory and HDD in which KVM overtakes VMware. Like [29], the author used a simple test case in which he used XP as a guest OS.

S. Varette et al. [15] evaluated energy-efficiency of VMware ESXi 5, KVM 0.12 and Xen 4.0, using NUMA architecture using HPC implementation. The authors used HPL benchmarking tool and the Grid 5000 platform to investigate the performance of different hypervisors in a well-regulated and similar to HPC

environment. The authors concluded that there is a sustainable performance impact introduced by the virtualization layer across all types of hypervisors.

Hwang et al. [4] investigated open source and commercial hypervisors (Hyper-V 2008R2, vSphere 5.0, KVM 2.6.32-279, and Xen 4.1.2). The authors stated that there is no impact by increasing the number of virtual CPUs on performance from one vCPU up to four vCPUs on all hypervisor. In our work, we will show that there is a high impact of vCPU on performance.

Graniszewski et al. [17] evaluated open source and commercial hypervisors (Hyper-V, ESXi, OVM, Virtual Box, and Xen) like Hwang et al. [4] investigated, but in [17] the authors used the latest version of each hypervisor in CMP architecture. Moreover, they included type II hypervisor in their experiments to show the difference as well. More important, they criticized the work of M. Kedziora [42] by stating that "Since 2012 the VMware is the market leader, followed by Microsoft". Graniszewski et al. [17] recommended that Xen is the best choice for the small size organization while VMware and Microsoft Hyper-V are good for enterprise but they are more expensive than Xen. However, the authors did not investigate the effect of vCPU-VM and vCPU-pCPU for hyper threading enabled NUMA architecture for each hypervisor.

Sogand et al. [43] compared the performance of three hypervisors: VMware ESXi 5.0, Xen 3.0.13 and KVM using two quad core CPUs with hyperthreading enabled (i.e., total 16 logical CPU) and 4x146 GB Hard Drive. As a guest OS, the authors installed RedHat Enterprise Linux 6.2 on each hypervisor. According

to their measurements, no single hypervisor has the best performance for CPU utilization, disk utilization, response time, and downtime: VMware and KVM performed better in terms of application response time and CPU utilization as compare with Xen, but Xen performed well during live migration and showed less downtime than KVM and VMware. However, the authors did not show the actual work perform by each hypervisor during CPU utilization. Furthermore, they only conducted their experiments on two VMs running on each host machine with three different vCPU-VM configurations (6 vCPU, 12 vCPU, and 16 vCPU) such that they did not consider over allocation of vCPU-VM configuration. It should be noted that the latest state-of-the-art hypervisors supported to assign more than 16 vCPUs to VM [44]).

3.2 vCPU-pCPU Mapping and vCPU-VM Configuration

Zong et al. [19] analyzed the impact of the non-uniform virtual CPU and vCPU-pCPU mapping on CMP system using Xen 3.4.0 hypervisor. The authors investigated the impact of non-uniform vCPU-pCPU mapping by running multithreaded applications (Apahe and TPC-H). The authors concluded that both dynamic and static non-uniform vCPU-pCPU mapping have the same performance implication i.e., negative impacts. Furthermore, the authors found out that the application performance instability such as throughput has a linear relationship with the de-

gree of non-uniformity. For instant, the performance will be more stable if each pCPU have the same number of vCPU. However, The authors did not test and analyze the impact of vCPU-pCPU mapping on NUMA system and hyper threading enabled architecture.

Yuxia and Wnzhi [45] investigated the impact of VMs scheduling on multicore systems. They applied different VM-mapping combinations using KVM hypervisor and NAS benchmark. They concluded that if the number of threads increases (i.e., number of threads more than pCPUs), then the scheduler used in KVM hypervisor can not work effectively in multicore NUMA architecture due to the NUMA unaware property of CFS algorithm used in KVM. In addition, the authors in [45] found out that due to the dynamic behavior of co-running VMs, the overall performance and VMs scheduling strategy will be affected. It should be noted that the latest hypervisors are NUMA aware [46, 47]. Therefore, the findings of [45] might not apply to the state-of-the-art hypervisors.

Sogand and Lars. [45] studied the performance implication of over allocation of vCPUs to VMs at NUMA architecture, using VMware ESXi 5.5 hypervisor. They observed that the performance could decrease up to 20% when there is a massive over-allocation of physical CPU resources. Moreover, the best performance was gained when there were only a few VMs with no over-allocation. Such work is different from our work. Authors tested different hypervisor than us. We tested Citrex xenServer and KVM hypervisors. Also, the authors did not investigate vCPU-pCPU mapping.

The summary of related work is shown in Table 3.1. As a summary of our related work, we leave the reader with Table 3.1 to compare the work done by each paper. As we see in Table 3.1, most of the authors analyzed and compared various hypervisors without investigating the effect of vCPU-VM configuration and vCPU-pCPU mapping. However, some of them analyzed either vCPU-VM configuration or vCPU-pCPU mapping by using a certain hypervisor.

In this thesis, prior to evaluate the hypervisors (Citrix xenServer and KVM), the best vCPU-VM configuration as well as vCPU-pCPU mapping strategies is analyzed for each hypervisor using NUMA architecture. In addition, the effects of hypervisor, virtual machine allocation, vCPU-VM configuration, and vCPU-pCPU mapping on performance in terms of CPU utilization are investigated.

Table 3.1: Summary of related work

Work	Hypervisor Used	Architecture	Virtualization Type	Benchmarking Tool	Hypervisor Comparison	vCPU-VM Configuration	vCPU-pCPU Mapping
Charles David [28]	KVM, Xen	CMP	HAV, PV	PTS	Yes	Randomly	Randomly
Babu et al [29]	OpenVZ, Xen, xenServer	CMP	Container, PV, FV	Unixbench	Yes	Randomly	Randomly
Mancas [30]	Vmware, KVM	CMP	Not Mentioned	Passmask	Yes	Randomly	Randomly
S. Varette et al, [15]	Hyper-V, KVM, VMware, VirtualBox	NUMA	FV, HAV	HPL	Yes	Randomly	Randomly
Hwang et al. [4]	VMware, Xen Hyper-V, KVM	SMP	HAV	Ramspeed, Netperf	Yes	Randomly	Randomly
Zong et al. [19]	Xen 3.4.0	CMP	PV	Apache, TPC-H	No	Randomly	Not Randomly
Yuxia and Wnzhi [45]	KVM	NUMA	PV	NAS	No	Randomly	Not Randomly
Sogand Lars [21]	VMware EXSi 5.5	NUMA	Not Mentioned	Vcenter, Performance, SSH + Sar	No	Not Randomly	Randomly

CHAPTER 4

RESEARCH METHODOLOGY AND EXPERIMENTAL DESIGN

In this chapter, we introduce our proposed methodology as well as our experimental design. Subsection 4.1 starts with some research questions that will influence our research methodology and our experimental setup. In subsection 4.2 and 4.3, we discuss our proposed methodology and methodology architecture. We discuss experimental design in subsection 4.4.

4.1 Research Questions

To properly design and conduct an evaluation environment, an answer to some research questions must be found. These questions are listed as follows:

1. What is the effect of hypervisors such as Citrix xenServer and KVM on CPU utilization?
2. What is the effect of virtualization technology on performance?
 - (a) What is the effect of VM on performance?
 - (b) What is the effect of vCPU on performance?
 - (c) What is the effect of vCPU-pCPU pinning strategies on performance?

4.2 Proposed Methodology

In our evaluation process, we want to follow a systematic problem-solving approach [48, 49] allowing us to scientifically deliver our contributions explained in section 1.4. In this study, empirical method [48, 49] is used which is based on evidence, observations, and experiments.

As shown in Figure 4.1, the proposed methodology consists of six steps. In the first step, we presented our research objectives. In the second step, we conducted a literature review for our research objectives, to investigate the effect of the virtualization layer, vCPUs-VMs allocation, and vCPUs-pCPUs matching on performance. In step three, we built an experimental design and selected an appropriate approach to achieved our research objectives. Performance evaluation is presented in step four. In step five, we calculated statistics from experiments. Finally, in step six, we analyzed and drew meaningful conclusions from our work.

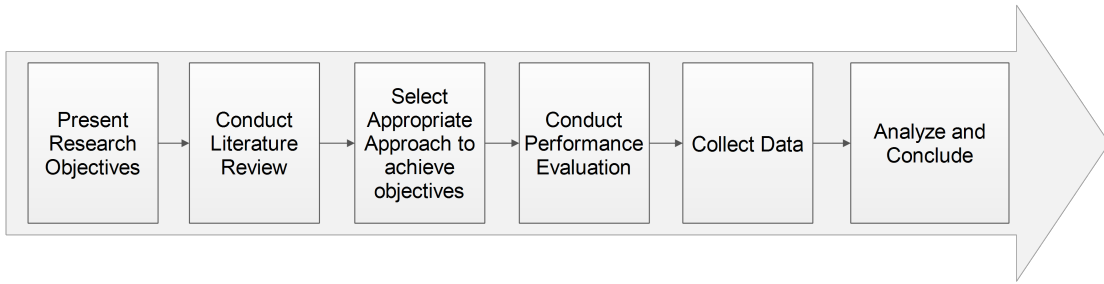


Figure 4.1: Proposed Methodology



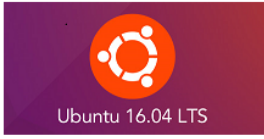
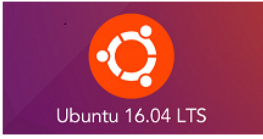




Benchmarking Software		
Guest Operating System <hr/> Virtual Machine		
Type I Hypervisor		
Physical Machines		

Figure 4.2: Proposed Methodology Architecture

4.3 Methodology Architecture

Our architecture, as shown in Figure 4.2, starts with two identical physical servers. The specification of these two servers is shown in Table 4.1. We installed two hypervisors (Citrix xenServer and KVM) on the physical servers. The two physical machine have similar architecture and specifications in order to achieve a fair assessment. Next, we created virtual machines running Ubuntu 16.04 as a guest OS. We built virtual machines on each hypervisor in order to provide the test environment. Finally, the PTS benchmarking tool is installed on each virtual machine as a traffic generator and analyzer

4.4 Experimental Design

Various experiments were conducted for CPU utilization using two state-of-the-art hypervisors. Before starting our main experiments, we investigated the effect of virtualization layer on elapsed time. It should be mentioned that we tested only CPU bound operations. In order to investigate and choose the best vCPUs-VMs allocation and mapping strategy for better CPU utilization and maximum performance. We focused our test measurement for under allocation (i.e., the number of vCPUs less than available logical CPUs), balance allocation of vCPUs to VMs (i.e., equally divided available logical CPUs among VMs), and over-allocation (the number of vCPUs more than available logical CPUs) and vCPU-pCPU mapping strategies.

PTS benchmarking tool is used to generate the workload and analyze the re-

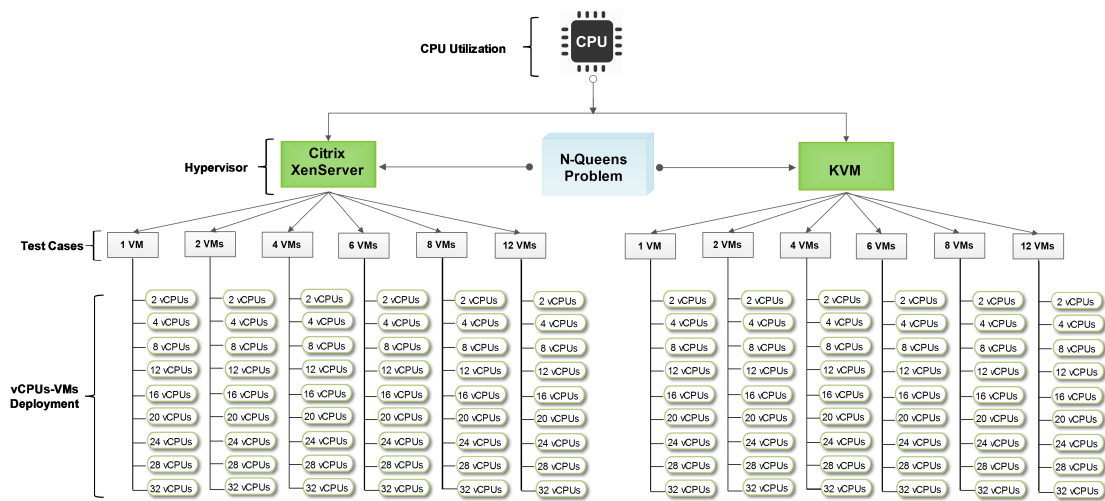


Figure 4.3: Experimental Design

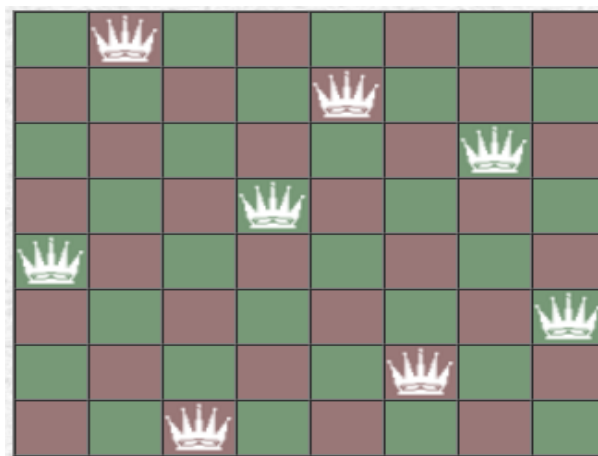


Figure 4.4: One Solution of N-Queens Problem (8 x 8 board) [5]

sults and a detailed evaluation is conducted to measure elapsed time and CPU utilization. PTS contains a variety of test profiles. For CPU bound operations, we chose two important test profiles called N-Queens and John-the-Ripper benchmarks. Workloads were generated using N-Queens benchmark, which report the elapsed time in seconds. We also measured the CPU utilization at hypervisor level for both hypervisors using Linux Top command.

4.4.1 N-Queens Benchmark

N-Queens is an open-source OpenMP benchmarking tool [27] that solves the N-Queens problem. N-Queens problem is a classical combinatorial problem, widely used as a benchmarking tool by researchers for CPU-intensive calculation that have different workloads and simple structure [50, 51]. The problem involves placing N queens on an N x N chessboard such that no queen can attack any other. Thus, a solution requires that no two queens share the same row, column, or diagonal. It is also used to test how the various hypervisors perform under calculation intensive operations [50, 51]. The N-Queens problem sizes are shown

Table 4.1: Specification of the Servers

Specifications	Server 1	Server 2
Hardware Model	Intel Xeon	Intel Xeon
Processor Speed	2 GHz	2 GHz
CPU Processor	12 Cores	12 Cores
Logical Processors	24 cores	24 cores
Main Memory (RAM)	64 GB	64 GB
Storage Capacity	1024 GB	1024 GB

Table 4.2: N-Queens problem with elapsed time (seconds)

Board Size (N x N chessboard)	Number of Solutions to N Queens Problem	Without Virtu- alization	Using xenServer	Using KVM
1	1	<0	<0	<0
2	0	<0	<0	<0
2	0	<0	<0	<0
4	2	<0	<0	<0
5	10	<0	<0	<0
6	4	<0	<0	<0
7	40	<0	<0	<0
8	92	<0	<0	<0
9	352	<0	<0	<0
10	724	<0	<0	<0
11	2680	<0	<0	<0
12	14200	<0	<0	<0
13	73712	<0	<0	<0
14	365596	<0	<0	<0
15	2279184	<0	<0	<0
16	14772512	<0	<0	<0
17	95815104	3.38	3.46	3.56
18	666090624	23.28	23.56	23.75
19	4968057848	191.5	193.72	201.18

in Table 4.2. As the problem size increases, the corresponding possible solutions and the elapsed time to solve the problem also increasing. For a regular-sized board (8 x 8), there are 92 distinct solutions, one of them is shown in Figure 4.4. In this thesis, we tested each hypervisor for different queens size ranges from 4 to 19.

4.4.2 Test Configurations

This evaluation is composed of two main experiments; Citrix xenServer-based setup and KVM-based setup as shown in Figure 4.3. We have four main factors in our experimental design namely: type of hypervisor, VMs, vCPUs, and workload. The objectives of these test configurations are to investigate the effects of hypervisor, VM, vCPU, and pinning strategies on performance. For every experiment setup, six test cases and nine vCPU-VM configurations are presented. Figure 4.3 illustrates the experimental design and the details of each test configuration. In Figure 4.3, there are two different hypervisors, each hypervisor has six different test cases. Each test case has nine different deployments i.e., allocation of vCPUs to VMs. In every deployment, we run N-Queens benchmark for sixteen different workloads (i.e., total $2 \times 6 \times 9 \times 16 = 1728$ experiments).

Each experiment is conducted on an identical separate server. Therefore, all the hardware resources of the server are fully dedicated for each hypervisor and the results obtained are fairly and reliably analyzed.

4.4.3 Test Cases

In these test cases, we gradually increase the number of active VMs on the top of each hypervisor and vary the number of vCPUs inside VM as well as the workload. The aim is to examine how the number of deployed load on a system and different vCPUs-VMs configuration will effect the overall performance in terms of CPU utilization and elapsed time.

One Virtual Machine

To evaluate the effect of hyperviosr, VMs, and vCPUs on performance, only one VM is running by having different vCPUs configuration. We focused on our test measurement for under allocation, balance, and over-allocation of vCPU-VM configuration to investigate the effect of each factor on performance.

We started our test configuration with the simple case; one VM has two vCPUs using N-Queens benchmark for different workload i.e., problem size range from 4 to 19. We measured average CPU utilization in percentage at VM level the elapsed time for each N-Queens problem size. Then, we gradually increased the number of vCPUs in the same VM (i.e., 4, 8, 12, 16, 20, 24, 28, and 32 vCPUs) until we investigated the best vCPU-VM configuration for which maximum performance was obtained in terms of CPU utilization and elapsed time. In addition to this, to investigate the effect of virtulization layer for CPU bound operations, we run one VM on the top of both hyperviosrs and allocate all physical CPU resource to one VMs (e.g., one VM having 24 vCPUs).

Two Virtual Machines

To further analyze the effect of VMs and vCPUs on performance, we doubled the number of active VMs and performed nine different experiments (vCPUs-VMs deployment) like we did in ‘One Virtual Machine’ test case.

Four Virtual Machines

In this case the number of active VMs are doubled than case two (total active VMs are four). The vCPUs-VM configurations are the same as for case one.

Six Virtual Machines

In this case, the number of active VMs are six, the rest of the configurations are the same as for case one.

Eight Virtual Machines

In this case, the number of active VMs are doubled (i.e., three times of case two and two time of case four), the other configurations are the same to as case one. Moreover, to investigate the effect of pinning strategies and the importance of over allocation of computing resources, we performed four different experiments using eight concurrent VMs. The aim was to observed the significance of over allocation of vCPUs and select the best vCPU-pCPU mapping strategies for better CPU utilization.

Twelve Virtual Machines

In this case, the number of active VMs are twelve. The rest of configurations remains the same as for case one.

CHAPTER 5

RESULTS AND DISCUSSION

In this chapter, the results have been discussed that were obtained using the PTS benchmarking tool. The objectives of these experiments are to investigate the effects of hypervisor, VM, vCPU, and pinning strategies on performance. The experimental results are shown in Figures 5.1 - 5.11 and Tables 5.1 - 5.3.

5.1 The Effect of Virtualization Technology

This test is designed to investigate the effect of virtualization technology layer (hypervisor) on performance in terms of CPU bound operations. For this test, only one VM allocated 24 vCPUs, having Ubuntu 16.04 as a guest OS, running at the top of both hypervisors, as well as a host OS on a bare-metal (non virtualized) machine. This test is performed using two powerful servers, server specifications are shown in Table 4.1. In this experiment, the N-Queens benchmark is used as a stress test to judge the virtualization overhead for CPU bound operations. The performance (elapsed time) of non virtualized machine against commercial and

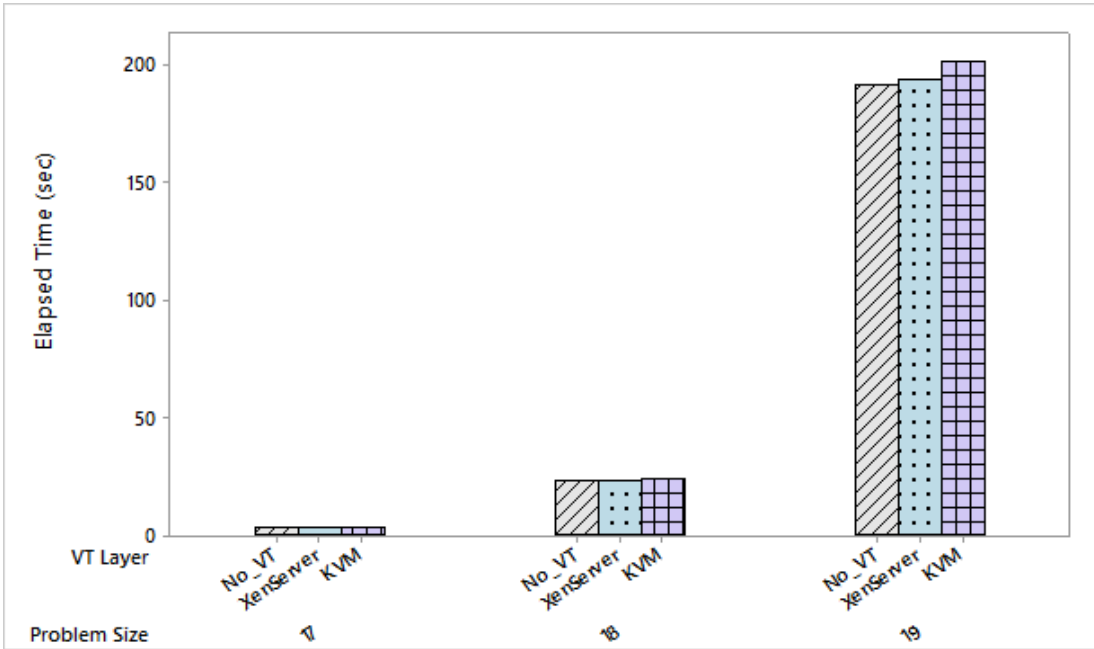


Figure 5.1: The effect of Virtualization Layer using N-Queens benchmark

open source hypervisor are given in Figures 5.1 and 5.2. Figure 5.1 illustrates the effect of virtualization on performance using three different workloads (17, 18 and 19; which is low, medium, and high workload) of N-Queens benchmark. The results illustrate that for low and medium workload there is no significant performance overhead but for heavy workload a low performance overhead is observed i.e., performance is decrease by 0.6% and 5% using Citrix xenServer and KVM respectively. One of the reason of performance reduction for heavy workload is that when we used heavy workload there are more context switching due to high elapsed time (i.e., CPU cycle are wasted instead of being utilized by vCPUs) and NUMA processor affinity between vCPUs as compared to low and medium workload.

To ensure that the CPU utilization (i.e., elapsed time) seen with N-Queens benchmark was not an anomaly, John-the-Ripper benchmark (CPU bound oper-

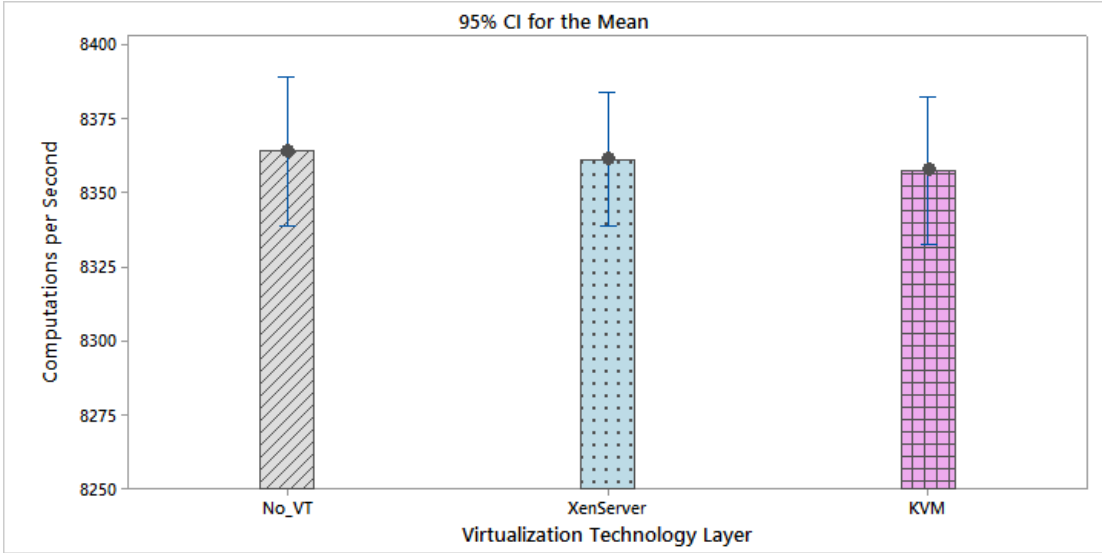


Figure 5.2: The effect of Virtualization Layer using John-the-Ripper benchmark (John-the-ripper) is used as a benchmark with the same settings. Figure 5.2 shows the result of Jon-the-ripper. Both benchmarks were run six times, and the results were averaged. Both results illustrated that for CPU bound operations using only one VM and consuming all CPUs, the virtualization overhead is almost minimal.

$$E = t(\alpha/2) * s/\sqrt{n} \tag{5.1}$$

Both benchmarking results are statistically verified using equation 5.1. Where E is the maximum error with one degree of confidence, (alpha) using two tail distribution, s is the standard deviation, and n is the number of samples. Confidence Interval (CI) with 95% significance level and p value is calculated. CI levels overlap with each other and p value is less than 0.05. So, the results are significant and there are no significant differences among the results.

5.2 The Effect of Virtual Machines on Performance

To investigate the effect of VMs on performance. In these experiments, we vary the number of VMs running on the top Citrix xenServer and KVM hypervisor. We also vary the number of vCPUs allocation to VMs as well as the workloads, as already discussed in chapter 4. For both hypervisors total 1728 ($2 \times 16 \times 6 \times 9$) observations were obtained where (2) is the number of hypervisors used in our experiments, (16) represents workloads, (6) shows different test cases of VMs running on top of each hypervisor, and (9) represents different vCPU-VM configurations. Out of 1728 obtained observations, the significant observations ($324 = 3 \times 2 \times 6 \times 9$) were found significant for CPU utilization as shown in Tables 5.1 and 5.3. Based on the possible solutions and elapsed time, we chose problem size 17 (low), 18 (medium) and 19 (heavy) workload.

Furthermore, for each experiment, the average CPU utilization in percentage at hypervisor level is measured. But, only CPU utilization in percentage is insufficient to investigate the effect of VMs on performance, especially when the CPU utilization level is 100% . Then, we can not judge the effect of VMs on performance. Therefore, the elapsed time to solve the N-Queens problem was calculated to trace how much actual work is performed by CPU. In the remaining of the thesis, we focused on elapsed time instead of CPU utilization in percentage.

To systematically investigate the effect of VMs on performance, we performed three main experiments: under allocation, balance allocation, and over alloca-

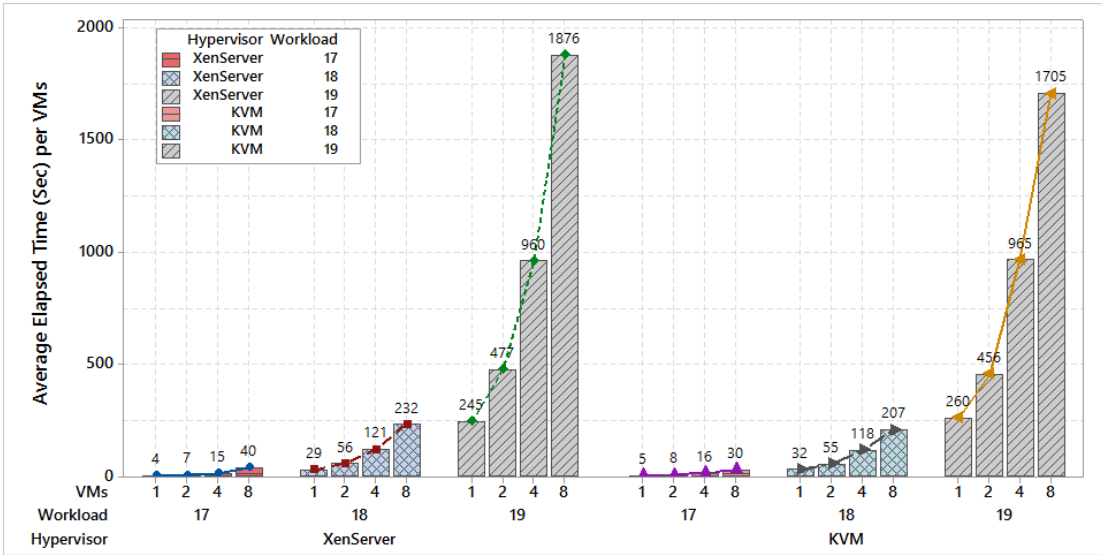


Figure 5.3: The effect of Virtual Machines on Performance - Under allocation

tion of computing resources. We discussed the details of each experiment in the following subsections.

5.2.1 Under Allocation

Under allocation is the case in which the total number of vCPUs assigned to VMs are less than the total logical CPUs. Based on user configuration for under allocation, the physical CPU cores are not fully utilized as shown in Figures 5.9 and 5.10. As a result, poor performance in terms elapsed time in seconds can be expected.

Figure 5.3 shows the effect of VMs on performance for under allocation of computing resource on two different hypervisors. In this experiment, the number of vCPUs are kept fixed (i.e., total 16 out of 24 vCPUs are allocated to VMs), while the number of VMs and workload are increased from one VM to eight VMs and low workload to high workload, respectively. As the number of concurrent

VMs increases, the elapsed time in seconds also increases i.e., nearly double for each workload. In a cloud environment, under allocation is not recommended due to the poor performance in terms of CPU utilization. Moreover, we can not run many VMs per our need, because the performance of the system decreases with increases in number of VMs. In order to achieved better performance (i.e., low elapsed time to solved N-Queens problem) and maximum CPU utilization, we need to consume all CPUs. Therefor, we carried out experiments and discussed in subsection 5.2.2 for balance allocation, where all the CPU cores were allocated to active VMs.

5.2.2 Balance Allocation

The maximum performance (i.e., low elapsed time) can be achieved if 100% CPU is utilized (24 out of 24 CPU logical CPUs are utilized), which is one of the main objectives of cloud computing. In balance allocation, physical CPU is utilized 100% as shown in Figures 5.9, 5.10. Figure 5.4 depicts the effect of VMs on performance using Citrix xenServer and KVM hypervisors for balance allocation. In balance allocation, the number of vCPUs were kept fixed, while the number of workload and active VMs varied. In this experiment, 24 out of 24 logical CPUs were equally divided among VMs, such as: 24 CPUs cores were assigned to one VM; 12 vCPUs, 6 vCPUs, 4 vCPUs, 3 vCPUs and 2 vCPUs were allocated to other five test cases.

Figure 5.4 illustrates the effect of VMs on performance using balance alloca-

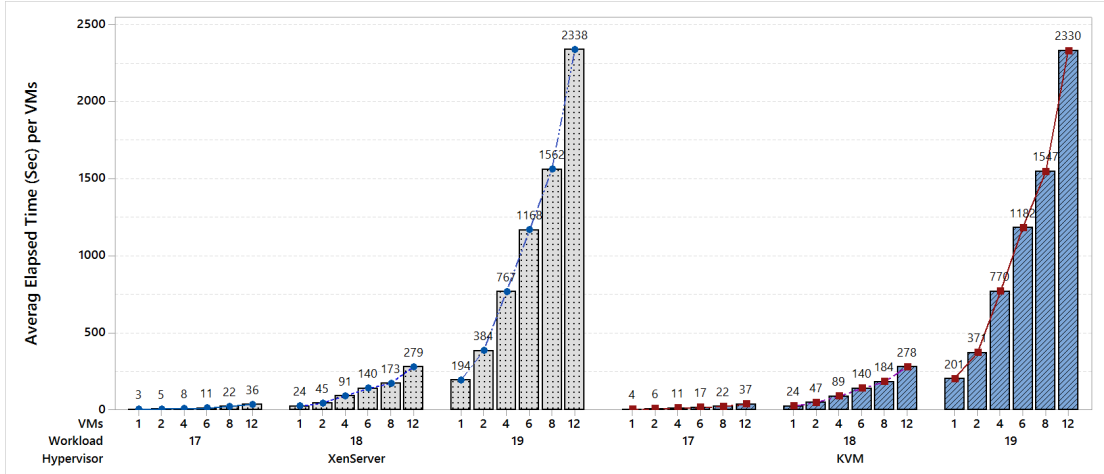


Figure 5.4: The effect of Virtual Machines on Performance - Balance allocation. By comparing the results of balance allocation with under allocation, in every test case and workloads, the improved performance (better elapsed time) was achieved as shown in Figure 5.6.

If we compare the average elapsed time of each test case, inside balance allocation test configuration (1 VM with 2 VMs; 2 VMs with 4 VM; 4 VMs with 8 VMs) using any workload, the average elapsed time is almost double, although the CPU utilization level is 100%. In one VMs test case, total available physical CPU resources (24 out of 24 logical CPUs) are allocated to one VMs. Therefore the total elapsed time is minimum. For two VMs test case, the VMs time share the CPU resources such as 50% CPU is be used by VM1 and 50% is used by VM2, therefore the elapsed time is almost double (98% increase) by comparing with one VM test case and so on.

We used linear regression model [52] for balance allocation to predict the future elapsed time (i.e., if more or less than 12 VMs are running) using the following formulas:

$$Y = \beta_0 + \beta_1 * VM \quad (5.2)$$

$$\beta_0 = \frac{(\sum Y) * (\sum VM^2) - (\sum VM) * (\sum VM * Y)}{n * (\sum VM^2) - (\sum VM)^2} \quad (5.3)$$

$$\beta_1 = \frac{n * (\sum VM * Y) - (\sum VM) * (\sum Y)}{n * (\sum VM^2) - (\sum VM)^2} \quad (5.4)$$

Where Y is the dependent variable (elapsed time) and plotted it on Y-axis. VM is the independent variable and plotted at on the X-axis which is the number of active VMs running on the top of hypervisor. Beta 1 is the slope of the line and Beta 0 is the Y-intercept [53]. Using equation 2, elapsed time can be predicted from active VMs by the following equations:

$$ElapsedTime(Sec)_{Citrix} = -4.8 + 195.45 * activeVMs \pm \epsilon, R^2 = 0.99 \quad (5.5)$$

$$ElapsedTime(Sec)_{KVM} = -10.3 + 195.45 * activeVMs \pm \epsilon, R^2 = 0.99 \quad (5.6)$$

Both prediction models shown in equation 5.5 and 5.6 give a strong prediction for new observations due to the R square values which are very close to 1. We also calculated the CI considering the significant level as 95% for each workload

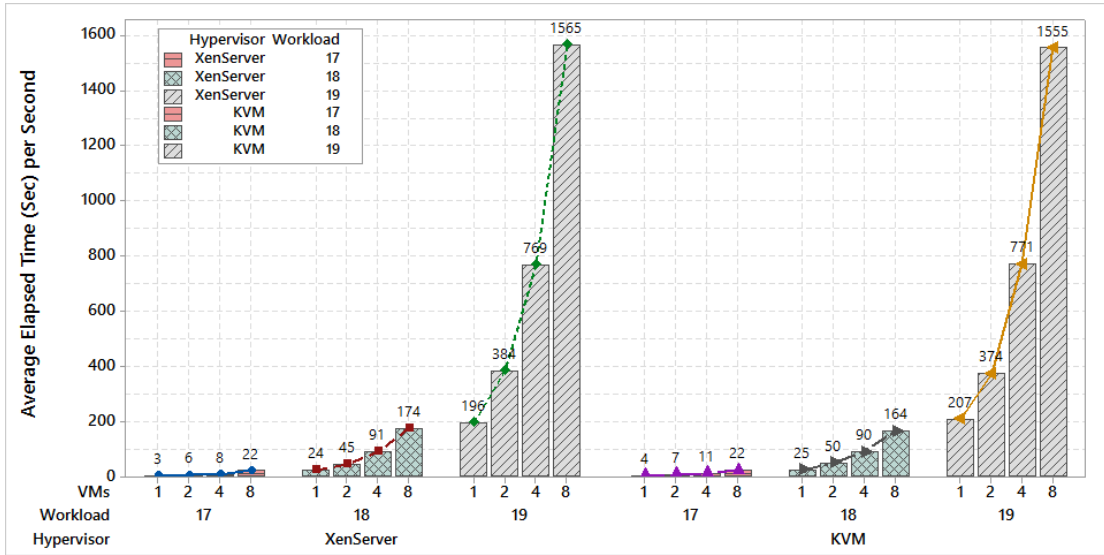


Figure 5.5: The effect of Virtual Machines on Performance - Over allocation

and for 6 test cases, namely: 1 VMs, 2 VMs, 4 VMs, 6 VMs, 8 VMs and 12 VMs.

It is clear that all CIs levels and means do not overlap with each other. This is an indication of decrease in performance by increasing the number of concurrent VMs from one to N, where N is the number concurrent VMs.

Figure 5.4 also illustrates the comparison between Citrix xenServer and KVM hypervisor for three different workload. It's clearly shown that for each workloads no significant difference was identified in either of the hypervisors.

5.2.3 Over Allocation

Figure 5.5 illustrates the effect of VMs on performance using over allocation (i.e., total 32 vCPUs are allocated to VMs). However, Figure 5.6 depicts a comparison among all test cases (under, balance, and over allocation). In case of under allocation, 16 vCPUs were assigned to each VM. In addition, in case of balance allocation, 24 vCPUs were allocated to each VM, 32 vCPUs were configured per

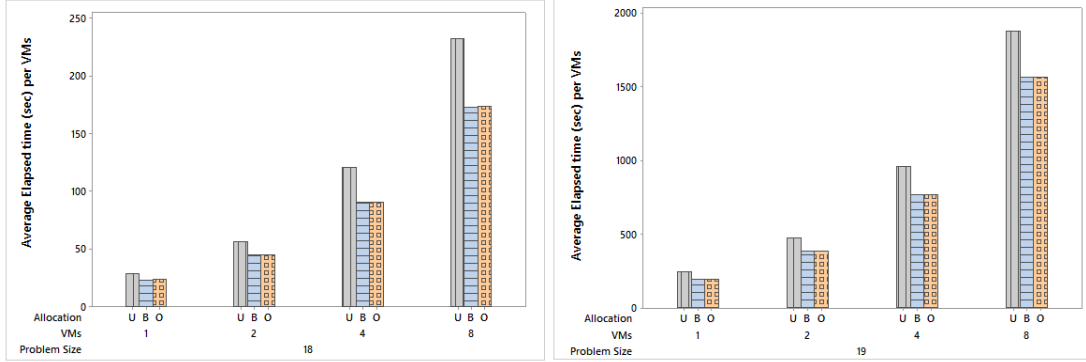


Figure 5.6: The effect of Virtual Machines on Performance

VM in case of over allocation. The performance (elapsed time) of over allocation and balance allocation were found analogous for two different workload. However, anomalous behavior of lowest performance (elapsed time) was shown by under allocation results for medium and heavy workloads. Therefore, it is needed to allocate a suitable vCPUs for each VM for better performance. The significance of vCPU-VM allocation will be discussed in details in subsection 5.3.

5.3 The Effect of Virtual CPUs on Performance

The cloud service providers are interested to know how much resources (vCPUs-VMs) should be allocated for maximum performance. Since large number of VMs are running in cloud environment and sharing physical computer resources, a risk of poor performance arises due to over allocation of physical CPU resources. These performance bottlenecks should be investigated, quantified, and avoided.

Here, we are testing the impact of vCPU assigned to VM. Previous studies [21, 43] showed that the system performance could be affected by using different ways of the pCPUs. Each virtual machine is configured with a number of vCPUs.

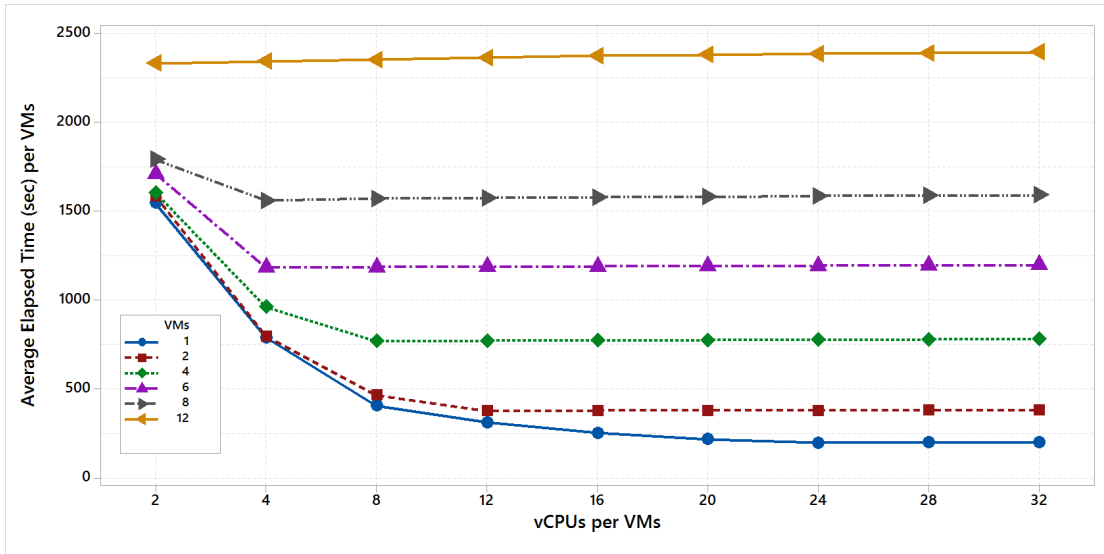


Figure 5.7: The effect of Virtual CPUs on Performance - Citrix xenServer

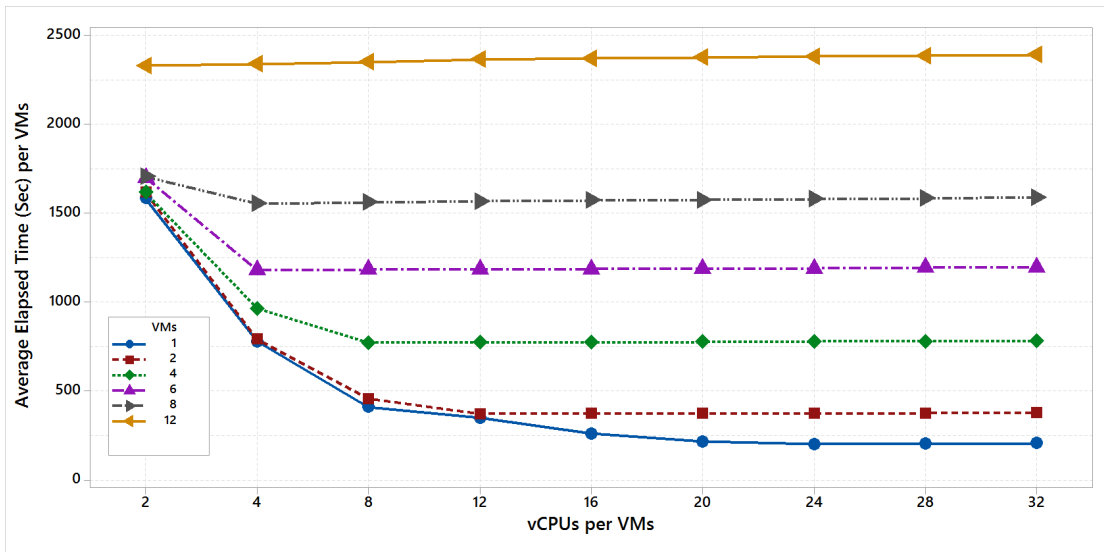


Figure 5.8: The effect of Virtual CPUs on Performance - KVM

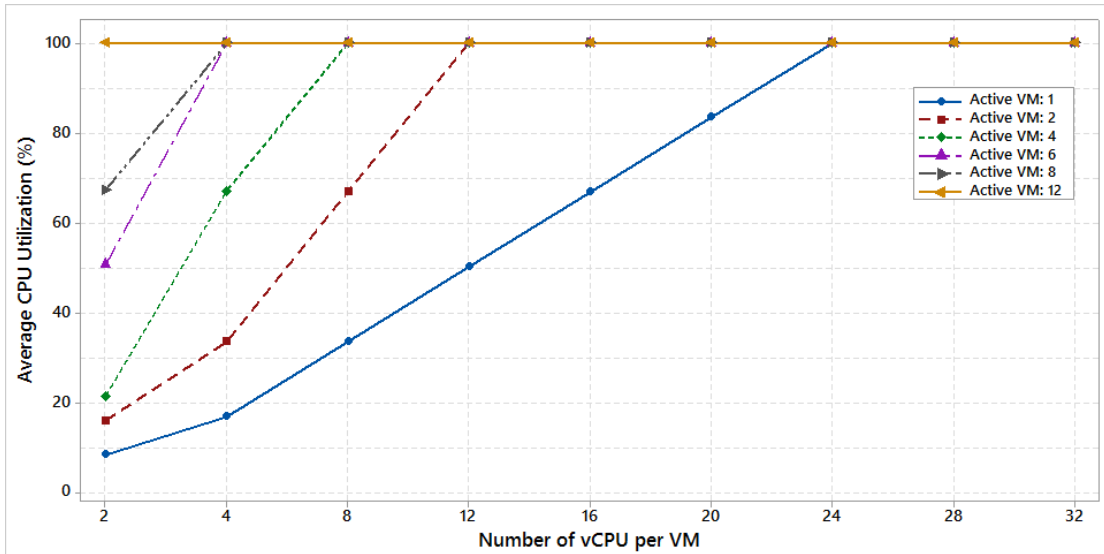


Figure 5.9: Total CPU Utilization at Hypervisor Level - Citrix xenServer

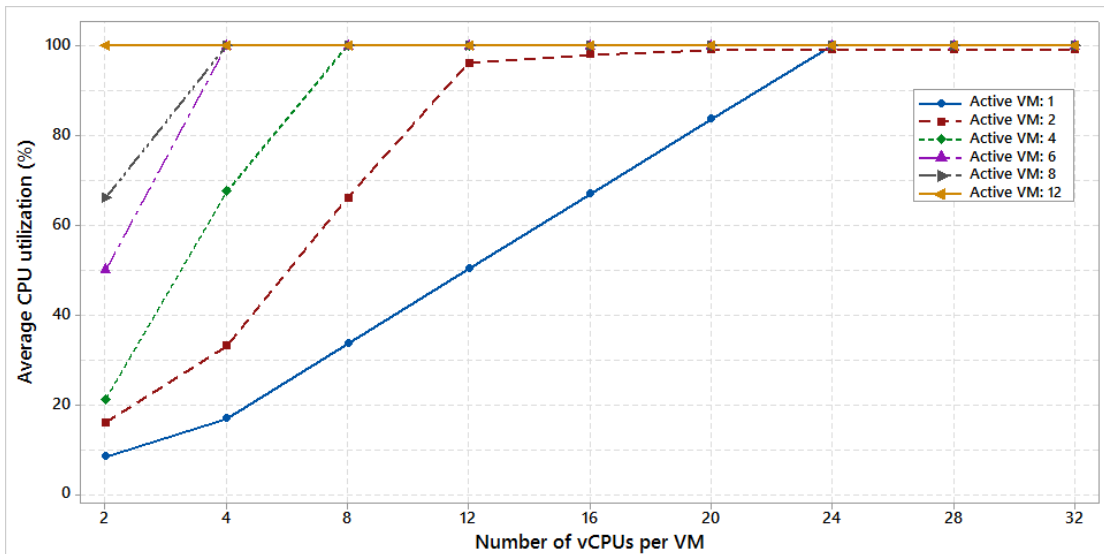


Figure 5.10: Total CPU Utilization at Hypervisor Level - KVM

The performance of a VMs having eight vCPUs will be doubled as compared to four vCPU-VM configuration (e.g., balance and under allocation). One can decide to use available pCPUs in two opposite ways such as by using few VMs having large number of vCPUs, or large number of VMs having small number of vCPU.

Figures 5.7 and 5.8 show the elapsed time in second for different vCPUs configurations (vi.e, CPU-VM configuration are 2, 4, 8, 12, 16, 20, 24, 28, and 32 vCPUs) using heavy workload (problem size 19). In both figures, the number of active VMs increases from 1 to 12 with nine different vCPUs-VM configuration. From both figures, it is clearly seen that, the performance is improved by allocating more vCPUs to VMs. However, there is a performance threshold for vCPUs i.e, 24 out of 24 logical CPUs are consume. After the threshold, no significance improvement was observed. The similar trend was observed in the data presented in Tables 5.1 and Table 5.2, where the performance was decreased after crossing the threshold. In One VM test case, the effect of over allocation of vCPUs to VMs was low, but it was significantly high for other test cases. The time sharing of CPU resources by VMs, in case of over allocation, could be the possible reason. That why, there was no or very small time sharing in balance allocation. Time sharing increases the number of context switches among VMs. The overhead, due to excessive context switching between VMs and NUMA processor affinity, will result performance reduction, and CPU cycle will be wasted instead of being utilized by the VMs. After further analysis, higher performance implication of over allocation was observed in 6, 8, and 12 VMs tests cases as compare to the other

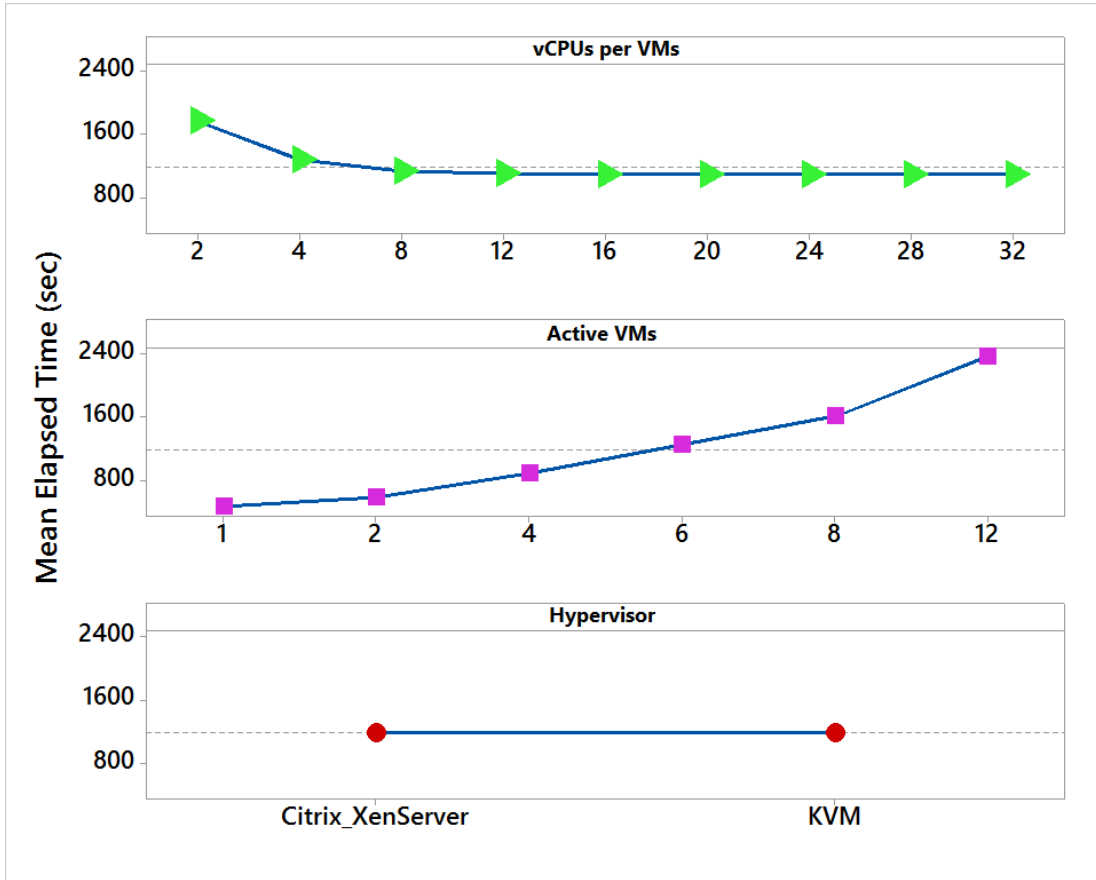


Figure 5.11: Mean Elapsed time of Hypervisor, VMs, and vCPUs

test case (1,2, and 4). In addition to this, the CPU utilization for Citrix xenServer and KVM are given in Figures 5.9 and 5.10. Both figures show CPU utilization in percentage at hypervisor level i.e., how much the VMs are using the physical CPU resources.

5.3.1 Concluding Remarks

In this section, we summarized the effect of hypervisors, VMs, and vCPUs on elapsed time. Figure 5.11 illustrates the mean of elapsed time in seconds for two hypervisors, active VMs, and vCPUs assigned to VMs using heavy workload. The

results reveal that mean elapsed time of both hypervisors are similar for active VMs as well as for vCPUs per hypervisor. So, for CPU bound operations, there is no significance difference to use commercial or open source hypervisor. The results also depict that by increasing the number of active VMs from one to twelve VMs, the performance will be decreases by 1106% i.e., elapsed time to solve the N-Queens problem increased. In addition, it is clearly seen that, the performance is improved by allocating more vCPUs to VMs. However, there is a performance threshold for vCPUs (i.e., balance allocation of vCPUs). After the threshold, no significance improvement was observed.

5.4 The Significance of Over Allocation

The over allocations of vCPUs to VMs is also important in a cloud environment. If a cloud service provider did not use over allocation of vCPUs, they may not be able to use all the physical cores after live migration or idle VMs. For instance, there is one host with 24 CPU cores, and there are two VMs running on the host and each VM has 12 vCPUs. If one VM is migrated to another host, crashed, or become idle then twelve physical CPU cores will not be used, although one VM is overloaded. However, if each VM configure more than 24 vCPUs then there would be enough vCPUs to utilize by overloaded VM after live migration or idle VM.

To highlight the significance of over allocation of vCPUs on performance, we performed two more experiments namely: uniform vCPUs and non-uniform

vCPU-VMs configuration. In addition, we compared two pinning strategies based on the same experiments. The details are given in the following subsections.

Uniform vCPUs-VMs Allocation

In uniform vCPU configuration, each active VM has the same number of vCPUs but no over allocation. In this experiment, we vary the pinning strategies and fixed the number of VMs (eight VMs), vCPUs (i.e., each VM allocated three vCPU, total vCPUs = $8 \times 3 = 24$), and also fixed the number of workloads (low, medium, and high). Two out of eight VMs will run a low workload, two of them with medium workload while the remaining four VMs will run the heavy workload.

Figure 5.12 illustrates the uniform vCPU-VM configuration for two vCPU pinning strategies; pinning and no pinning strategy. In no pinning strategy, the hypervisor is free to schedule domain's vCPUs on any pCPUs. While in pinning strategy the hypervisor is free to schedule the Dom0 (hypervisor) vCPUs on any pCPUs and other active VMs' vCPUs are statically pinned to user define logical CPUs. We discussed the pros and cons of pinning and no pinning in section 5.5.

To investigate the effect of vCPUs on uniform vCPUs, we have run two different experiments with eight VMs each. The purpose of assigning different workloads to VMs, while keeping the same vCPUs configuration, was to investigate the effect of over allocation and pinning strategies. Figure 5.12 shows that after 26 and 28 seconds (depending on pinning strategies), the VMs having low workload become idle, due to low workload they finished their task early as compare to medium and high workload VMs. The other six VMs were still busy. However, the

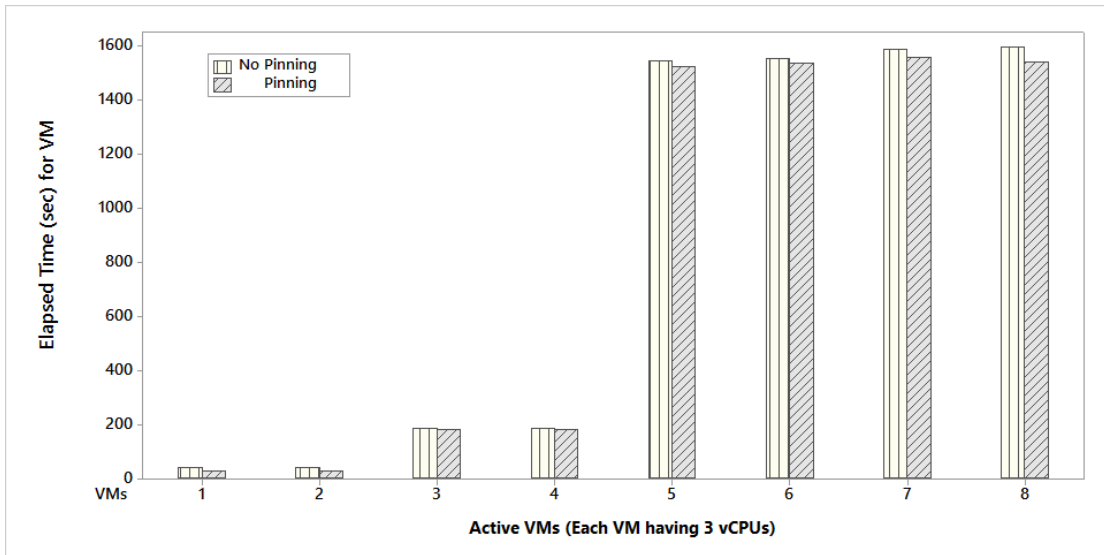


Figure 5.12: The effect of Virtual CPUs on Performance - Uniform vCPUs per VMs

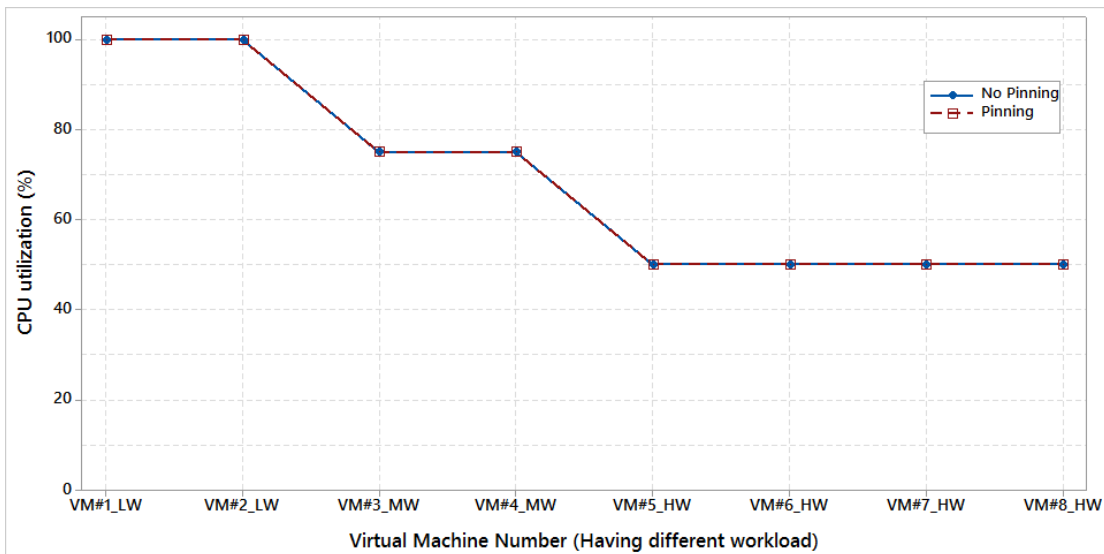


Figure 5.13: CPU Utilization for Uniform vCPUs

CPU utilization level dropped from 100% to 77% as shown in Figure 5.13. After 180 and 182 seconds (depending on pinning strategies), the VMs having medium workload become idle. Now four VMs out of eight VMs are idle while other four are still busy due to heavy workload. Thus, the average CPU utilization level drops to 52%. So, there was no significant difference among pinning strategies (pinning strategies have no effect on under allocation and balance allocation of vCPUs-VMs). For better CPU utilization, the optimum vCPU-VMs configuration is needed. In the next subsection, we will discuss the non-uniform vCPUs test configuration by which CPU utilization and performance (in terms of elapsed time) can be improved.

Non-uniform vCPUs-VMs Allocation

In non-uniform vCPU configuration, each active VM has the same vCPUs like uniform configuration, but we used over allocation. In this experiment, we vary the pinning strategies and fixed the number of vCPUs (6 vCPUs per VM), the number of VMs (8 active VMs), and the number of workloads (low, medium, and high). Two out of eight VMs will run a low workload, two of them with medium workload while the remaining four VMs will run the heavy workload. The aim of over allocation of vCPUs, in this experiment was to utilize all the physical cores after idleness of VMs. Because four VMs were busy due to high workload and other four VMs which having low and medium workload will be become idle due to low and medium workload.

Figure 5.14 shows the significance of over allocation of vCPUs and the result

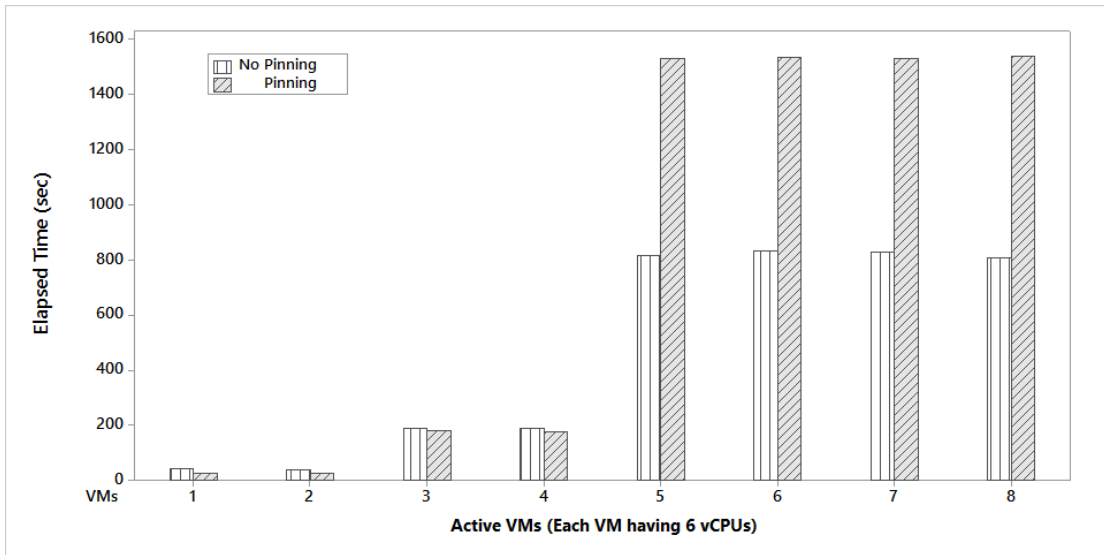


Figure 5.14: The effect of Virtual CPUs on Performance - Non-uniform vCPUs per VMs

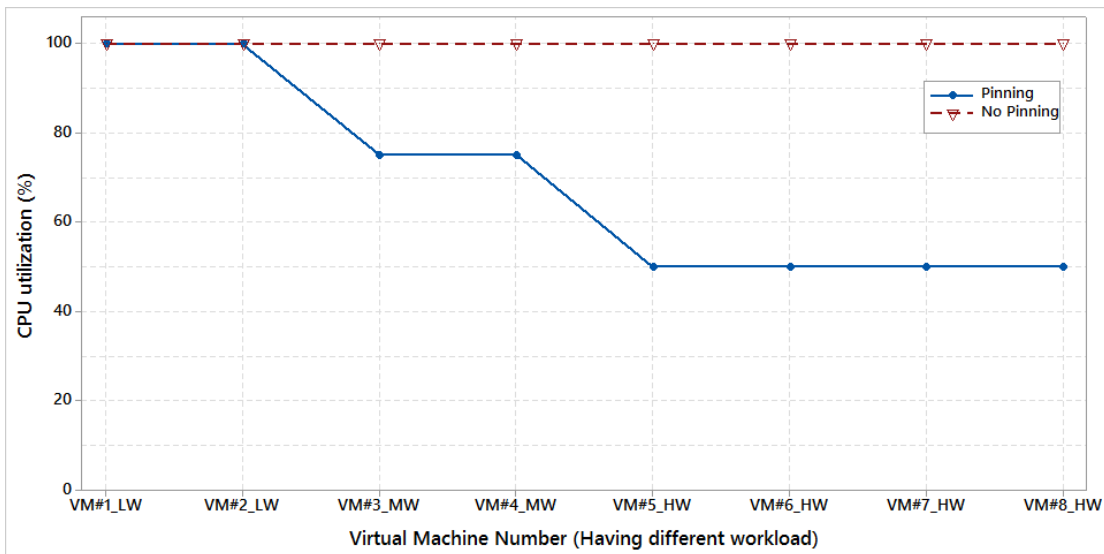


Figure 5.15: CPU Utilization for Non-uniform vCPUs

of two pinning strategies. It is shown, that after 28 and 26 seconds (depending on pinning strategies), the VMs having low workload became idle. The other six VMs having medium and high workload were still busy. But, this time the CPU utilization level for no pinning strategy did not drop from 100% as shown in Figure 5.15. Based on pinning strategies, the hypervisor may or may not assign the idle vCPUs of two idle VMs (having low workload) to VMs having the medium and high workload while no pinning strategy assign the idle vCPUs to medium and heavy workload VMs. After 180 and 178 seconds (depending on pinning strategies), the VMs having medium workload also became idle. Now, four out of eight VMs are idle while four are still busy due to the heavy workload. The no pinning strategy takes the advantage of over allocation of vCPUs, by using the free vCPUs. In addition, the CPU utilization level for no pinning strategy did not drop from 100% as shown in Figure 5.15. As a result, the performance (elapsed time) is improved and the elapsed time to solved the N-Queens problem for the heavy workload is minimized (i.e., almost half as compare with other two pinning strategies). While the pinning strategies can not utilize the free vCPUs, because it is restricted to run on particular vCPUs. In conclusion, pinning strategies have effect on over allocation (vCPUs-VMs) and better performance (elapsed time) and higher CPU utilization can be achieved using over allocation of vCPUs-VMs with no pinning strategy.

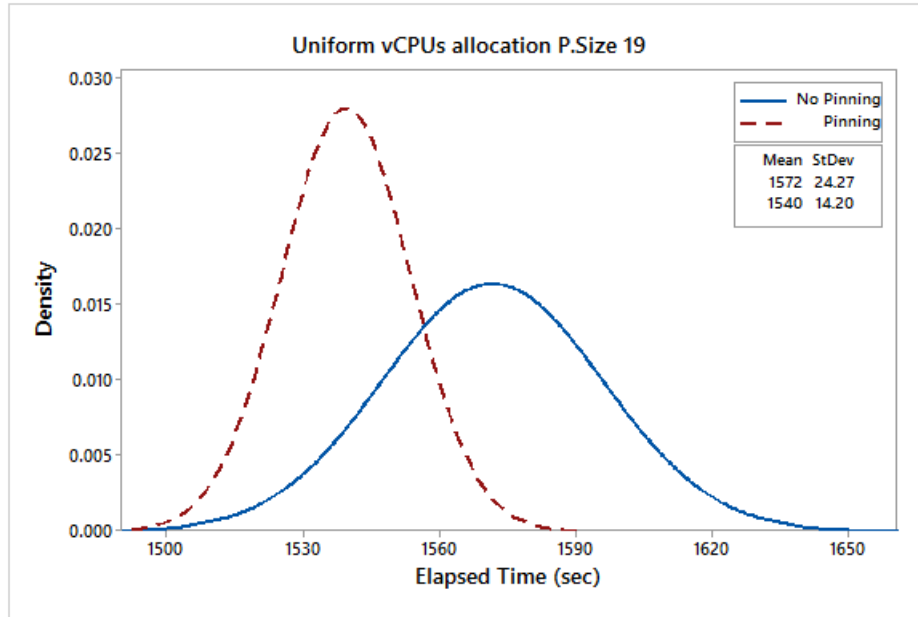


Figure 5.16: Normal distribution of Elapsed time (sec), Uniformed vCPU allocation

5.5 The Effect of Pinning Strategies on Performance

We already discussed the effects of pinning and no pinning on performance in subsection 5.4. Here, we briefly discuss the pros and cons of both strategies.

The effect of pinning strategies depends on workload types and vCPUs-pCPU allocation (under, balance, and over allocation). Figures 5.16 and 5.17 show normal distribution of elapsed time (sec) for two pinning strategies. For uniform vCPU allocation, it seems that both pinning strategies have similar normal distribution curve. But pinning strategy gives more stable performance for heavy workload i.e., mean and standard deviation of pinning strategy is 3%, 71% respectively lower than no pinning strategy as shown in Figure 5.16.

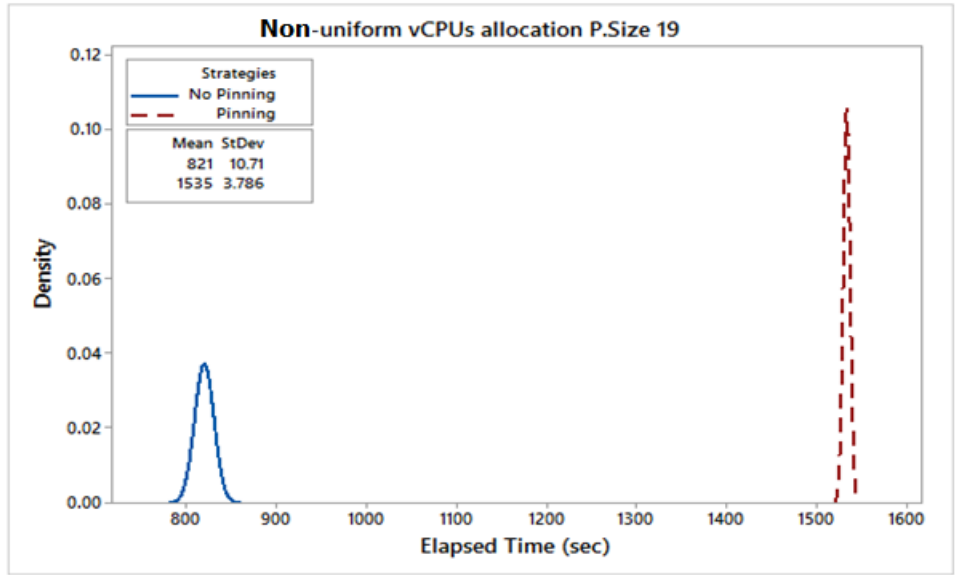


Figure 5.17: Normal distribution of Elapsed time (sec), Non-uniformed vCPU allocation

Nevertheless, the no pinning strategy having Non-uniformed vCPUs allocation performed better than pinning strategy as shown in 5.17. The elapsed time decrease by 74.78% for no pinning strategy.

Table 5.1: Elapsed time (sec) to solve N-Queens problem using Citrix xenServer

P. Size	Number of vCPUs	Active VMs: 1	Active VMs: 2	Active VMs: 4	Active VMs: 6	Active VMs: 8	Active VMs: 12
17	2	26.53	29.54	29.14	30.42	40.00	36.24
	4	13.59	15.06	15.36	10.78	21.90	37.67
	8	6.92	7.19	8.35	10.86	22.08	40.24
	12	4.86	5.09	8.44	10.97	22.89	40.66
	16	4.26	5.62	8.99	11.23	23.22	41.28
	20	3.81	5.81	9.15	11.87	23.78	42.22
	24	3.46	5.88	9.44	12.33	24.03	42.50
	28	3.49	5.99	9.23	13.23	25.76	43.20
	32	3.46	5.80	10.33	14.85	26.06	44.33
18	2	183.14	195.50	197.10	199.71	232.15	279.33
	4	109.57	97.49	120.51	145.66	173.88	282.73
	8	47.94	56.48	91.05	147.33	174.33	284.19
	12	33.28	44.90	91.67	147.98	176.56	288.30
	16	29.23	44.99	92.34	149.50	177.87	292.76
	20	26.04	45.15	92.99	152.87	178.33	294.73
	24	23.56	45.36	93.23	153.20	178.99	295.22
	28	23.60	46.33	94.08	154.89	179.82	296.61
	32	23.66	47.07	94.45	155.01	180.23	297.84
19	2	1509.53	1555.39	1591.24	1723.40	1876.00	2337.68
	4	800.55	803.41	959.75	1188.84	1565.33	2347.59
	8	402.55	476.67	768.80	1188.33	1579.87	2355.33
	12	273.76	383.88	768.99	1190.65	1582.51	2364.10
	16	244.91	383.97	771.22	1192.51	1584.76	2376.72
	20	218.16	385.19	773.74	1196.19	1587.09	2383.11
	24	193.72	385.33	775.11	1196.23	1591.72	2388.22
	28	194.06	386.48	777.61	1197.56	1593.02	2391.71
	32	195.7	387.58	779.15	1199.09	1595.01	2398.22

Table 5.2: Elapsed time (sec) to solve N-Queens problem using KVM

P. Size	Number of vCPUs	Active VMs: 1	Active VMs: 2	Active VMs: 4	Active VMs: 6	Active VMs: 8	Active VMs: 12
17	2	27.04	20.43	28.54	31.04	29.52	36.89
	4	13.81	14.12	16.49	16.93	22.33	38.29
	8	7.28	7.64	11.22	16.99	23.82	40.72
	12	6.58	6.45	11.54	17.23	23.97	40.83
	16	4.64	6.72	11.89	17.55	24.09	41.22
	20	3.91	6.78	12.23	18.23	24.98	42.12
	24	3.56	6.85	12.87	18.66	25.20	42.72
	28	3.68	7.09	12.54	19.20	25.33	43.80
	32	3.70	7.42	13.08	18.48	26.30	44.03
18	2	188.05	141.16	195.20	205.29	206.97	278.36
	4	94.27	96.03	118.00	139.81	164.43	283.06
	8	48.67	54.71	89.55	141.22	132.67	285.63
	12	46.18	47.39	90.56	141.49	149.56	289.98
	16	31.98	47.90	90.34	143.31	172.55	290.60
	20	26.33	48.64	91.23	145.98	156.80	291.66
	24	23.75	48.80	91.44	147.94	149.36	292.25
	28	24.58	49.25	92.78	149.66	131.84	294.30
	32	24.79	49.55	93.65	151.83	188.23	295.00
19	2	1584.86	1617.69	1616.63	1700.89	1705.34	2329.92
	4	778.25	792.54	964.91	1181.50	1554.88	2329.30
	8	410.80	456.05	770.89	1183.45	1562.76	2350.07
	12	349.86	371.08	772.98	1184.82	1568.32	2364.33
	16	260.38	373.60	774.23	1186.23	1572.22	2371.22
	20	216.20	374.75	777.87	1188.90	1574.22	2376.00
	24	201.18	374.04	778.02	1190.32	1579.85	2382.33
	28	203.99	375.03	779.33	1194.67	1584.34	2386.29
	32	207.39	377.05	780.55	1197.33	1589.69	2390.76

Table 5.3: Elapsed time (sec) to solve N-Queens problem using Citrix xenServer and KVM for balance allocation

Hypervisor	Workload	Active VMs	vCPU per VMs	Elapsed Time (Sec)
Citrix xenServer	17	1	24	3.46
		2	12	5.09
		4	6	10.77
		6	4	10.78
		8	3	21.77
		12	2	36.24
	18	1	24	23.56
		2	12	44.90
		4	6	90.93
		6	4	140.31
		8	3	172.74
		12	2	278.53
	19	1	24	193.72
		2	12	383.88
		4	6	767.30
		6	4	1188.84
		8	3	1561.92
		12	2	2337.68
KVM	17	1	24	3.56
		2	12	6.45
		4	6	11.07
		6	4	16.93
		8	3	22.24
		12	2	36.89
	18	1	24	23.75
		2	12	47.39
		4	6	89.49
		6	4	139.81
		8	3	183.62
		12	2	278.36
	19	1	24	201.18
		2	12	370.08
		4	6	770.32
		6	4	1181.50
		8	3	1547.05
		12	2	2329.92

CHAPTER 6

CONCLUSIONS AND FUTURE WORK

6.1 Conclusions

In this thesis, we tested the most recent hypervisors (open source and commercial) in terms of CPU utilization. This test was conducted in a cloud environment. Specifically, we analyzed the implication of virtualization technology layer, vCPU-VM assignment, and vCPU-pCPU mapping on performance in terms of CPU utilization. In addition, we proposed a suitable vCPUs configuration and pinning strategies for VMs in cloud environments. This configuration provides adequate performance in terms of elapsed time and CPU utilization. It should be mentioned that we tested only CPU bound operations. Cloud service providers and researchers will get the benefits when they deploy VMs in a cloud environment or evaluate open source and commercial hypervisors.

Using CPU bound operations, the results obtained from this evaluation showed that commercial (Citrix xenServer) and open source (KVM) hypervisors have similar performance in terms of elapsed and CPU utilization.

As per our observation, the performance of a system would degrade by running many VMs, improper allocation of vCPUs to VMs, or using unsuitable vCPUs-pCPUs pinning strategies. Moreover, we have found that elapsed time increases when there is a massive over allocation of vCPUs.

The experimental results revealed that the best performance (elapsed time) was gained when there was only a few active VMs with balance allocation and over allocation (i.e., over allocation using no pinning strategies). In addition, virtual CPUs to physical CPUs pinning strategies have no effect on performance for under allocation and balance allocation of vCPUs-VMs. However, better CPU utilization and low elapsed time were obtained using over allocation of vCPUs to VM with no pinning strategy.

We suggested that the cloud service providers and researchers should consider the effects of massive over allocation of vCPUs, VMs, and vCPUs-pCPUs mapping when they choose deployment strategies for better performance and best CPU resources allocation.

6.2 Future Work

Migration refers to moving a server environment from one place to another. With most virtualization solutions it is possible to move a virtual machine from one

physical machine to another. Migration is typically used to improve reliability and availability: in the case of hardware failure, the guest system can be moved to a healthy server with limited downtime, if any. It is also useful if a virtual machine needs to scale beyond the physical capabilities of the current host and must be relocated to physical hardware with better performance. CPU utilization during live migration of virtual machine from one host to another is one of the future directions for this work.

There are a variety of vendors for the virtualization environment and all of them claim that their virtualization hypervisor is the best, however they depend on the used application. The obtained results from our evaluation experiments can be validated using other commercial hypervisors (VMware, Hyper-V, and RHV). This will aid the cloud service providers in choosing which hypervisor to use for their specific needs.

In addition, CPU utilization for I/O bound operations has not been investigated. We used CPU bound operations and evaluated the effects of hypervisor, vCPU-VM configuration, and vCPU-pCPU mapping on performance. However, I/O bound operations or the combination of both could be used to get improved performance in terms of CPU utilization.

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