

A Fine-grained Performance-based Decision Model for Virtualization Application Solution



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Outline

- Introduction
- Architecture of VirtDM
- Metrics choosing
- VirtDM modeling
 - VirtDM formulation
 - VirtDM implementation
- Case study
- Conclusion

1.Introduction

- Datacenter virtualization problem
 - Virtualization Application Solution(VAS)
 - Decision of VAS for special virtualization application scenarios



- Systematic Decision method——VirtDM
 - Virtualization performance evaluation
 - MCDM: Multi-Criteria Decision-Making
 - Human preference
 - Comparable results

Datacenter virtualization problem



Application
Virtualization

Presentation
Virtualization

Desktop
Virtualization

Storage
Virtualization

Network
Virtualization

Vmware

Xen

KVM

Hyper-V

VirtualBox

.....

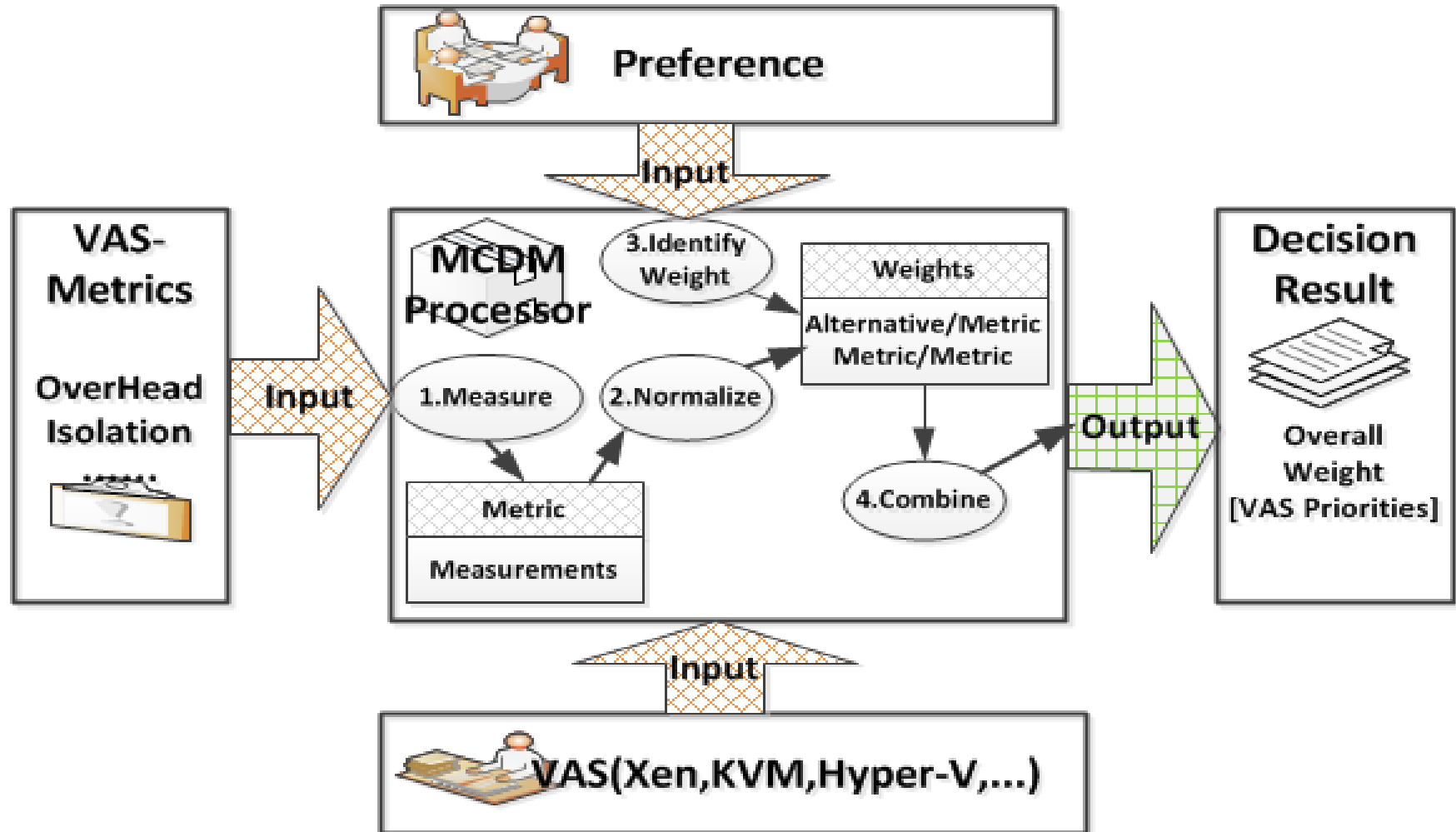


Problems of VirtDM

- ❑ What metrics should be taken into account to measured a VAS?
- ❑ How to quantify a datacenter architect's preference on these metrics?
- ❑ How to achieve an overall decision from different metrics' results and architects' preference?

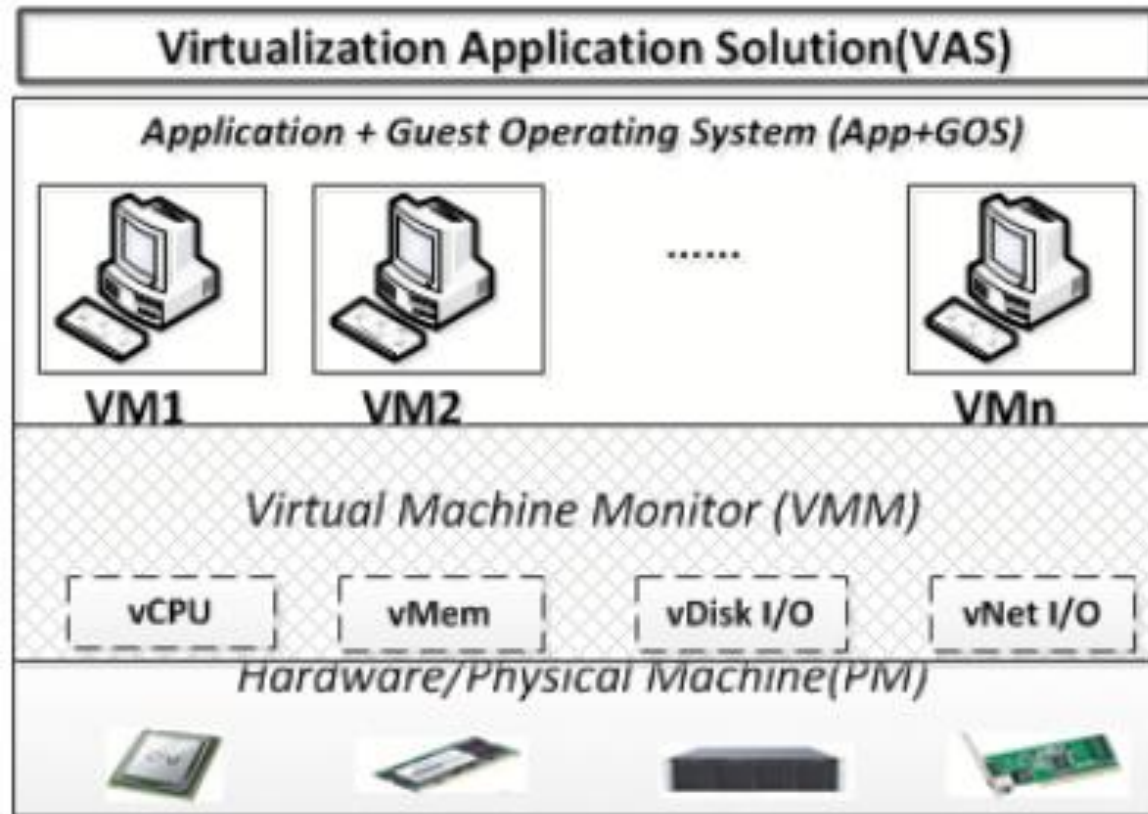


2. Architecture of VirtDM



What is VAS

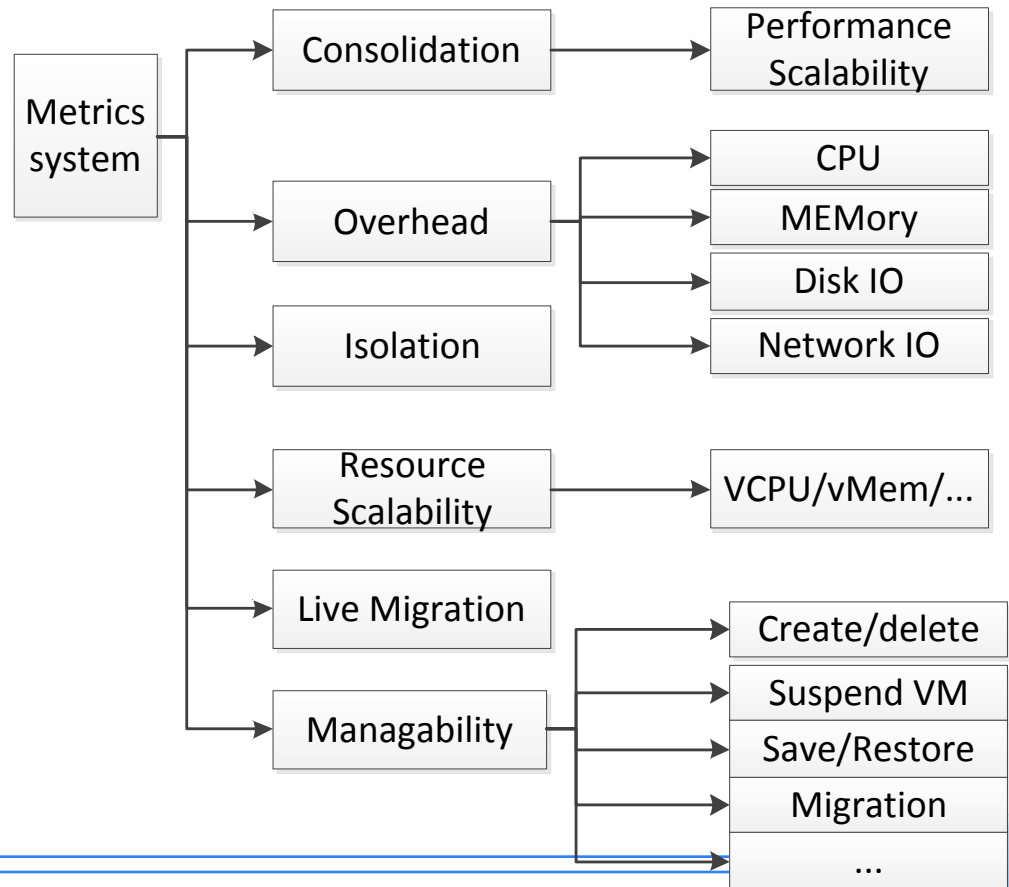
- Refers to a specific software and hardware implementation of virtualization technologies.
- Para-virtualization & Full virtualization
- Hardware with Intel VT or AMD SVM





3. Metrics Choosing

- Virtualization application solution
 - Virtualization Overhead
 - Manageability
 - Isolation
 - Consolidation
 - Live migration
 -





Virtualization Overhead

- Four workloads used to measure
 - CPU task, memory task, I/O task and Context Switch task.
- Performance degradation percentage is used to express the quantities of overhead



Manageability

- Include 5 metrics
- immeasurable metrics
 - VM resource scalability
 - Migration function
 - Consolidation functional scalability
- Measure with response time
 - VM snapshot save/resume efficiency
 - VM start/shutdown efficiency



Isolation

- ❑ run different stress tests - CPU bomb, memory bomb, I/O bomb
- ❑ cause extreme resource consumption and refer their VMs as bad VMs
- ❑ measure the performance degradation of the normal workloads on a well-behaving VM.



Consolidation

- ❑ Uneasy to measure
- ❑ A good method is to use benchmark tool: SPECvirt sc2010.
- ❑ Scales the workloads on the System Under Test (SUT) until the SUT reaches its peak performance.



Migration

- Use Virt-LM benchmark
- It provides the results of four metrics
 - downtime
 - total migration time
 - the amount of migrated data
 - migration overhead



4. VirtDM Modeling

□ VirtDM Formulation

- Is a formulation for the VAS Multi-Criteria Decision-Making problem
- Main parts include: VAS candidates, Metrics, Decision-making Matrix

□ VirtDM Implementation

- Metrics quantification
- Metrics normalization with dimensionless method
- Weight identification \& pairwise comparison method
- AHP

Hierarchical structure of VirtDM

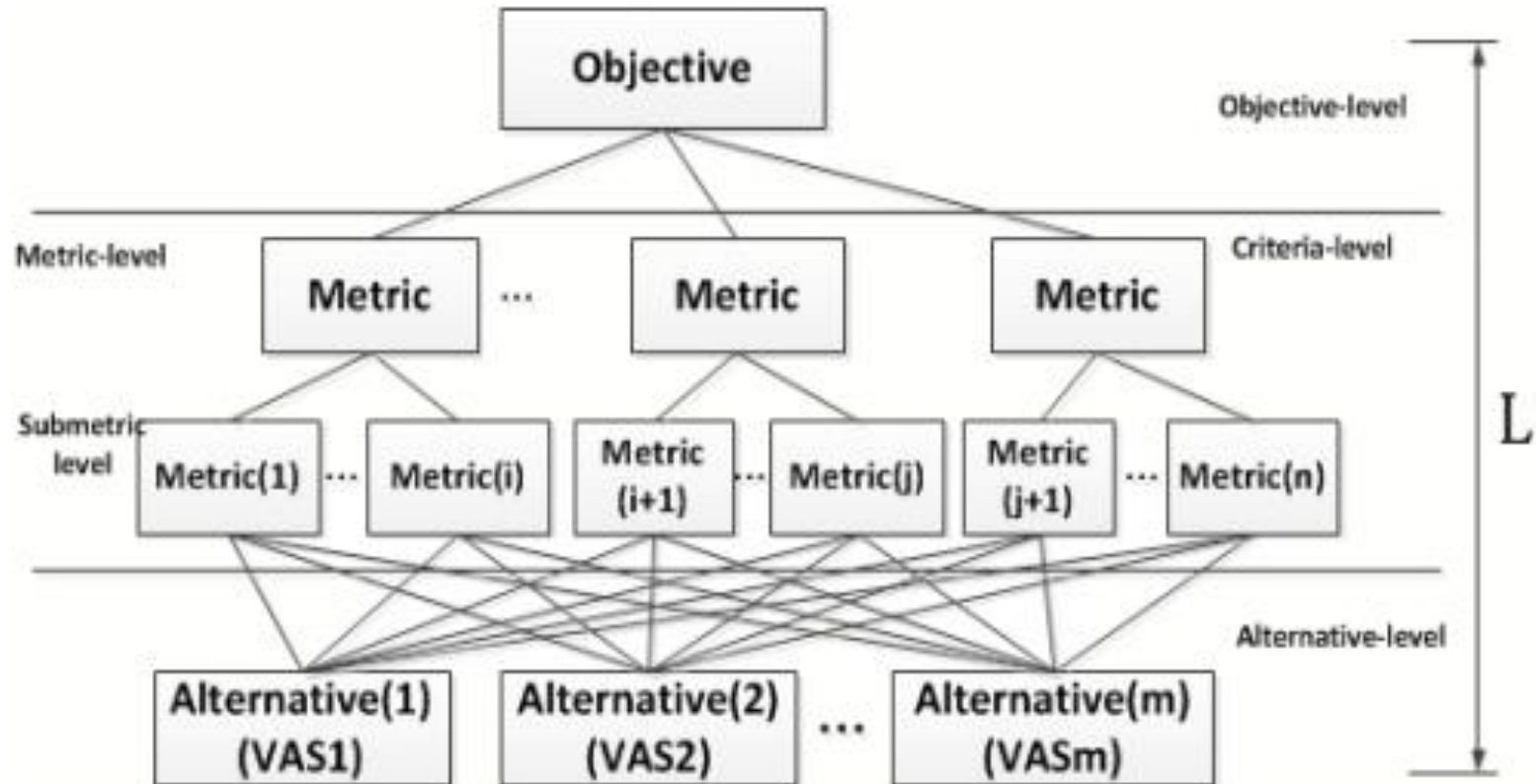


Fig. 3: The formulation of VirtDM with hierarchical structure



Problems

Problem 1. (Generalization problem). The MCDM problem of the VirtDM is provided with a hierarchy structure and must be decomposed into levels as shown in Fig. 3. It comprises L -levels ($L \geq 3$): alternative (VAS candidate, one fixed level), criteria (one or more of metric levels, apparently equal to $L - 2$ levels) as well as objective (decision objective level, one fixed level).

Problem 2. (Special case). Let $L = 4$ in problem 1, then we have a new hierarchy MCDM problem with four levels. In this way, in addition to the alternative-level and objective-level, the metric-level of general problem is divided into two levels: ML(metric-level) and SML(sub-metric level) as illustrated in Fig. 3. Each metric of the metric-level can be composed of several sub-metrics of the sub-metric level.



Definitions

Definition 1. (*Size of the hierarchy structure*). Assume problem 2 contain m alternatives (VASes), n sub-metrics and s metrics in the criteria level as well as 1 objective at the top level. Two adjacent levels are directly related.

If the i_{th} given metric of the metric-level contains n_i sub-metrics, it will satisfy the equation: $\sum_{i=1}^s n_i = n$.

Definition 2. (*Decision Attribute Matrix*). For each alternative, we can obtain a numerical value, called an attribute, for each metric of sub-metrics. Then, in problem 2 we have $m \times n$ attributes which comprise the decision basis of VirtDM. To store the decision attributes, we give a matrix: $D = (d_{ij})_{m \times n}$, ($i = 1, \dots, m, j = 1, \dots, n$), where the element d_{ij} represents the j^{th} sub-metric value of the i^{th} VAS alternative. This matrix is called Decision attribute matrix (DAM).

VirtDM implementation

Metrics quantification

Metrics normalization

Human Preference

Weight identification

Weight combination

- Metrics: quantitative and qualitative
- Normalization means:
 - vector normalization
 - linear scale transformation
 - (0-1) interval conversion

$$Score(A_i) = \sum_{j=1}^n r_{ij} * w_j. \quad (2)$$

$$V = R * W3 * W2 * W1 \quad (6)$$



Algorithm 1 Algorithm WIA : weight identification of pairwise comparison method.

Require:

- The amount of metrics for pairwise comparison, N ;
- The random index corresponding to N dimension, RI ;
- The pairwise comparison matrix, $P = (x_{ij})_{N \times N}$;

Ensure:

The weight vector of metrics, $W = (w_i)^T, i = 1..N$;

- 1: Determining and input the elements of P according to decision maker's preference by Satty's scale method [13].
- 2: Use geometric mean method as Formula (3): an approximate method to calculate the weight $W = (w_i)^T, i = 1..N$.

$$m_i = \prod_{j=1}^N x_{ij}, \quad \bar{w}_i = \sqrt[n]{m_i}, \quad w_i = \bar{w}_i / \sum_{i=1}^n \bar{w}_i. \quad (3)$$

- 3: Using Formula (4): an approximate method to calculate the maximum eigenvalue of P .

$$\lambda_{max} = \sum_{i=1}^n (\sum_{j=1}^n x_{ij} w_j) / n \cdot w_i. \quad (4)$$

- 4: Using Formula (5) to carry out CI and CR.

$$CI = (\lambda_{max} - n) / (n - 1), \quad CR = CI / RI. \quad (5)$$

- 5: If $CR < 0.5$ then goto step (6) to output the result weight vector: W ; else goto step (1);
- 6: **return** W .



5. Case Study

- ❑ Suppose a virtualized datacenter deployed preferring I/O performance
- ❑ Given and setup three VAS platforms environment for VirtDM implementation
- ❑ The purpose is to make decision the best VAS candidate

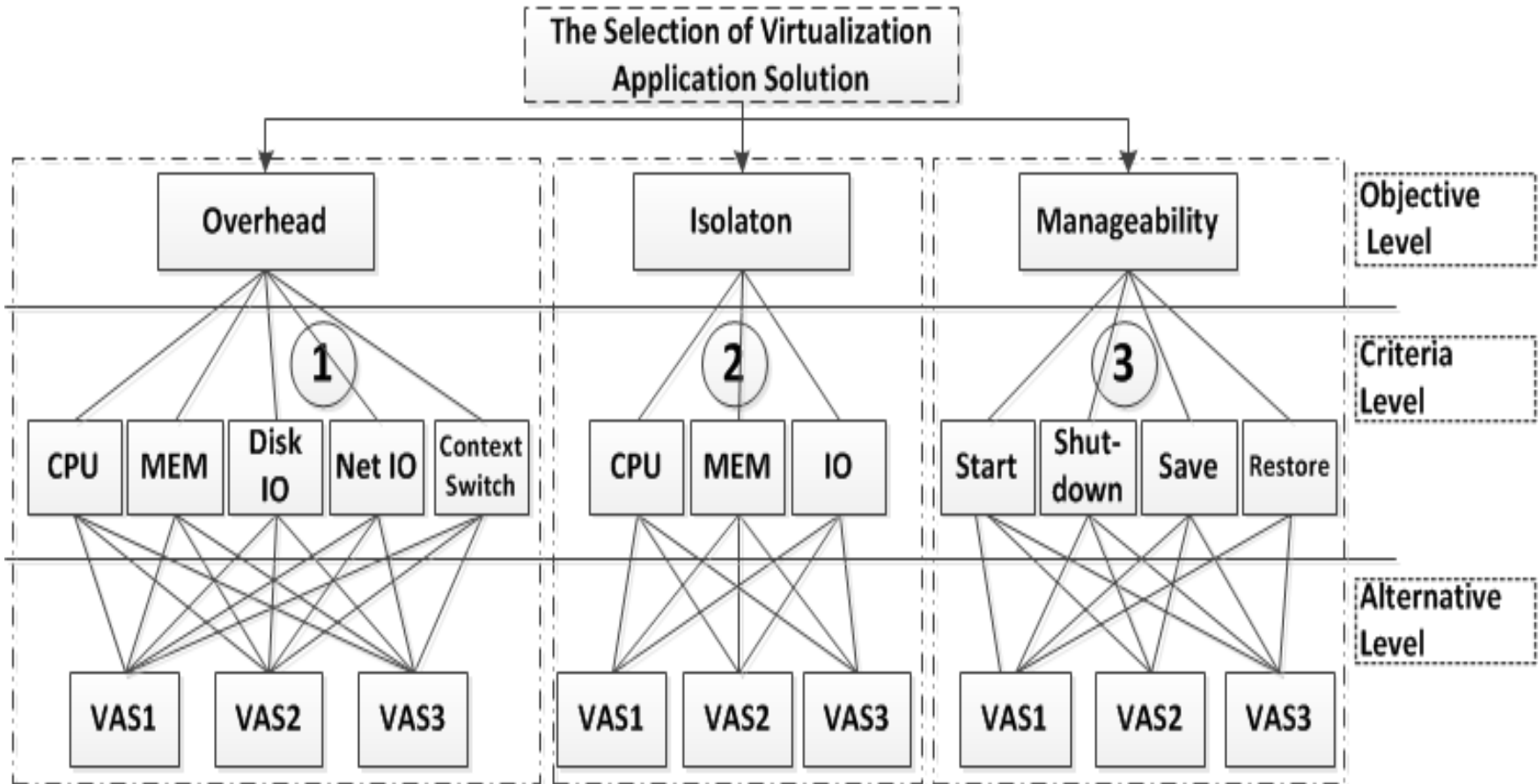
- ❑ Measuring the performance
- ❑ Overall decision process



Experimental Environment Setup

- 1) VAS-XEN-HV
 - Physical host is a Dell PowerEdge T710, with dual quadcore Intel Xeon processor E5620 at 2.4GHZ and 24GB of memory.
 - VMM is Xen-3.3.1 with Linux Kernel 2.6.18.8-xen
- 2) VAS-XEN-PV
 - Using the same host and VMM as VAS-XEN-HV but with a para-virtualized VM.
- 3) VAS-KVM
 - Using the same host and VM as VAS-XEN-HV but with a different VMM — KVM.

Hierarchical Metrics for decision



Performance Measurement

Table 1: Performance measurements from three VASes: XEN-HV, XEN-PV, KVM.

| | Overhead[%] | | | | | Isolation[%] | | | Manageability[sec] | | | |
|--------|-------------|------|----------|---------|------|--------------|------|------|--------------------|-------|------|-------|
| | CPU | Mem | Disk I/O | Net I/O | C.S. | CPU | Mem | I/O | start | shut. | save | rest. |
| XEN-HV | 8.1 | 15.5 | 51.1 | 7.7 | 41 | 0.6 | 35.7 | 42.4 | 21 | 2.4 | 17.9 | 16.2 |
| XEN-PV | 11.13 | 4.8 | 7.9 | 4.8 | 98 | 23.8 | 21.8 | 20.1 | 18.5 | 3.5 | 17.2 | 16.9 |
| KVM | 9.13 | 14.5 | 56.3 | 7.0 | 50 | 0.4 | 33.5 | 55.0 | 20 | 2.5 | 18 | 16.5 |

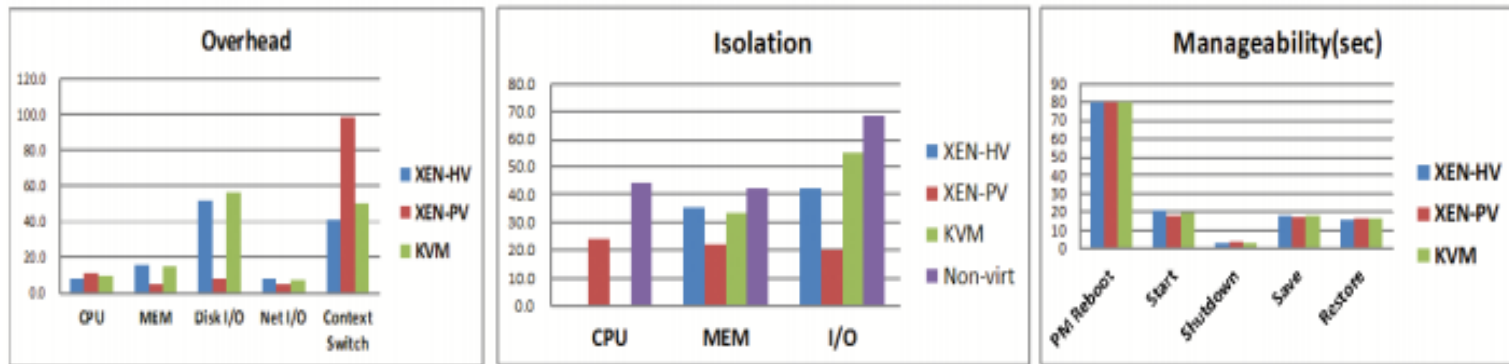


Fig. 5: The metrics results of three VASes



Overall Decision Process

Table 2: The normalized data

| | Overhead[%] | | | | | Isolation[%] | | | Manageability[sec.] | | | |
|---------|-------------|------|--------|-------|------|--------------|------|------|---------------------|-------|------|-------|
| | CPU | Mem | DiskIO | NetIO | C.S. | CPU | Mem | IO | Start | Shut. | Save | Rest. |
| XEN-HV | 0.38 | 0.18 | 0.12 | 0.27 | 0.31 | 0.38 | 0.19 | 0.26 | 0.28 | 0.30 | 0.28 | 0.31 |
| XEN-PV | 0.28 | 0.58 | 0.77 | 0.43 | 0.13 | 0.01 | 0.31 | 0.54 | 0.32 | 0.21 | 0.29 | 0.30 |
| Hyper-V | 0.34 | 0.25 | 0.11 | 0.30 | 0.56 | 0.61 | 0.50 | 0.20 | 0.40 | 0.49 | 0.42 | 0.39 |

- Give examples to demonstrate the usefulness of VirtDM
 - Performance data comes from measurement
 - Data is standardized and normalized
 - MCDM of VirtDM is used to get score for each VAS

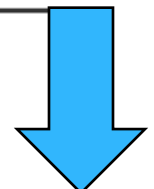
Constructing the decision-making matrix

□ Normalize the metrics

Table 2: The normalized data

| | Overhead[%] | | | | | Isolation[%] | | | Manageability[sec.] | | | |
|--------|-------------|------|------|------|---------|--------------|------|------|---------------------|-------|-------|------|
| | CPU | Mem | Disk | I/O | Net I/O | C.S. | CPU | Mem | I/O | Start | Shut. | Save |
| XEN-HV | 0.38 | 0.19 | 0.12 | 0.27 | 0.45 | 0.38 | 0.27 | 0.26 | 0.31 | 0.38 | 0.33 | 0.34 |
| XEN-PV | 0.28 | 0.61 | 0.77 | 0.43 | 0.19 | 0.01 | 0.44 | 0.54 | 0.36 | 0.26 | 0.34 | 0.33 |
| KVM | 0.34 | 0.20 | 0.11 | 0.30 | 0.36 | 0.61 | 0.29 | 0.20 | 0.33 | 0.36 | 0.33 | 0.33 |

□ Construct decision-making matrix



$$O = \begin{bmatrix} 0.38 & 0.19 & 0.12 & 0.27 & 0.45 \\ 0.28 & 0.61 & 0.77 & 0.43 & 0.19 \\ 0.34 & 0.20 & 0.11 & 0.30 & 0.36 \end{bmatrix}, I = \begin{bmatrix} 0.38 & 0.27 & 0.26 \\ 0.01 & 0.44 & 0.54 \\ 0.61 & 0.29 & 0.20 \end{bmatrix}, M = \begin{bmatrix} 0.31 & 0.38 & 0.33 & 0.34 \\ 0.36 & 0.26 & 0.34 & 0.33 \\ 0.33 & 0.36 & 0.33 & 0.33 \end{bmatrix}$$



Identifying weights based on preference

- AHP method is applied
- Pairwise comparison matrix

$$PO = \begin{matrix} & \begin{matrix} CPU & Mem & DiskI/O & NetI/O & Cont.S \end{matrix} \\ \begin{matrix} CPU \\ Mem \\ DiskI/O \\ NetI/O \\ Cont.S \end{matrix} & \begin{bmatrix} 1 & 1 & 0.111 & 0.14 & 0.333 \\ 1 & 1 & 0.143 & 0.125 & 0.2 \\ 9 & 7 & 1 & 0.5 & 2 \\ 7 & 8 & 2 & 1 & 3 \\ 3 & 5 & 0.5 & 0.333 & 1 \end{bmatrix} \end{matrix}, PI = \begin{matrix} & \begin{matrix} CPU & Mem & I/O \end{matrix} \\ \begin{matrix} CPU \\ Mem \\ I/O \end{matrix} & \begin{bmatrix} 1 & 3 & 5 \\ 0.33 & 1 & 3 \\ 0.2 & 0.33 & 1 \end{bmatrix} \end{matrix},$$

$$PM = \begin{matrix} & \begin{matrix} start & shut. & save & res. \end{matrix} \\ \begin{matrix} start \\ shut. \\ save \\ res. \end{matrix} & \begin{bmatrix} 1 & 1 & 0.17 & 0.17 \\ 1 & 1 & 0.2 & 0.2 \\ 6 & 5 & 1 & 1 \\ 6 & 5 & 1 & 1 \end{bmatrix} \end{matrix}, PP = \begin{matrix} & \begin{matrix} PCM1 & ove. & iso. & man. \end{matrix} \\ \begin{matrix} ove. \\ iso. \\ man. \end{matrix} & \begin{bmatrix} 1 & 3 & 9 \\ 0.33 & 1 & 4 \\ 0.11 & 0.25 & 1 \end{bmatrix} \end{matrix}$$



Combining weights

$$(1) W_1 = O \cdot W_o = \begin{bmatrix} 0.38 & 0.19 & 0.12 & 0.27 & 0.45 \\ 0.28 & 0.61 & 0.77 & 0.43 & 0.19 \\ 0.34 & 0.20 & 0.11 & 0.30 & 0.30 \end{bmatrix} \cdot \begin{bmatrix} 0.048 \\ 0.044 \\ 0.311 \\ 0.435 \\ 0.163 \end{bmatrix} = (0.254, 0.499, 0.248)^T$$

$$(2) W_2 = I \cdot W_i = \begin{bmatrix} 0.38 & 0.27 & 0.26 \\ 0.01 & 0.44 & 0.54 \\ 0.61 & 0.29 & 0.20 \end{bmatrix} \cdot \begin{bmatrix} 0.634 \\ 0.260 \\ 0.106 \end{bmatrix} = (0.338, 0.178, 0.484)^T,$$

$$(3) W_3 = M \cdot W_m = \begin{bmatrix} 0.31 & 0.38 & 0.33 & 0.34 \\ 0.36 & 0.26 & 0.34 & 0.33 \\ 0.33 & 0.36 & 0.33 & 0.33 \end{bmatrix} \cdot \begin{bmatrix} 0.075 \\ 0.081 \\ 0.422 \\ 0.422 \end{bmatrix} = (0.337, 0.330, 0.333)^T.$$

$$V = (W_1, W_2, W_3) \cdot W_p = \begin{bmatrix} 0.254 & 0.499 & 0.248 \\ 0.338 & 0.178 & 0.484 \\ 0.337 & 0.330 & 0.333 \end{bmatrix} \cdot \begin{bmatrix} 0.681 \\ 0.250 \\ 0.069 \end{bmatrix} = (0.281, 0.407, 0.313)^T.$$



7. CONCLUSIONS AND FUTURE WORK

- ❑ Design and implement the VirtDM model to serve the VAS decision making in a datacenter
- ❑ Provide a fine-grained, in-depth, and human friendly metrics system to cover essential performance characteristics of a VAS
- ❑ Many aspects of VirtDM are far from satisfying:
 - metrics system are to be improved
 - other MCDM methods excludes AHP method

Thank You!



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