

Quantitative Evaluation of Service Dependability in Shared Execution Environments

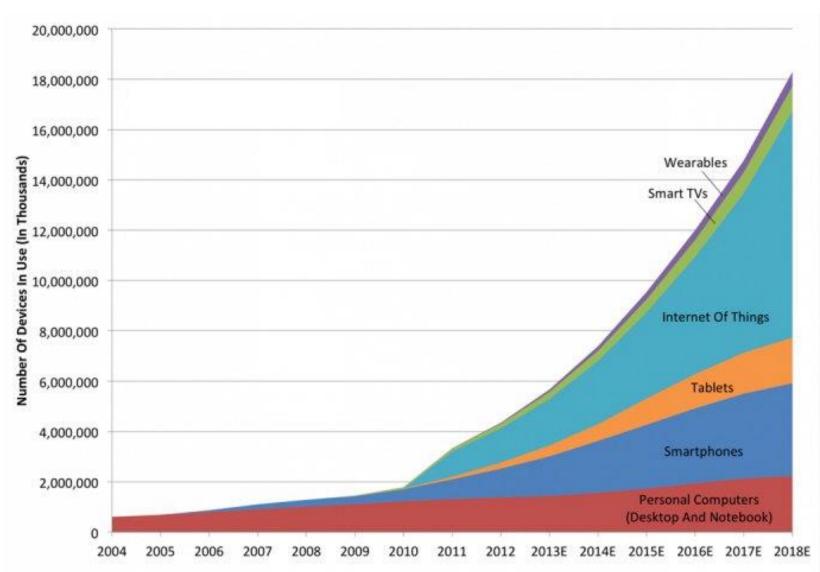
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Keynote talk, QEST 2014, Florence, Italy, Sept. 10, 2014



Explosion of IT Service Clients



Source: Gartner, IDC, Strategy Analytics, Machina Research, company filings, BII estimates



One Internet Minute



25,000+ new Apps added every month

Source: Intel, March 2012 (http://scoop.intel.com/what-happens-in-an-internet-minute)



Growing Data Centers



Maiden, North Carolina (Apple) 46 000 m²



Prineville, Oregon (Facebook) 28 000 m²



San Antonio (Microsoft) 43 000 m²



Chicago (Digital Realty) 100 000 m²

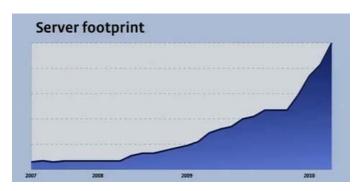


WU Growing Number of Servers

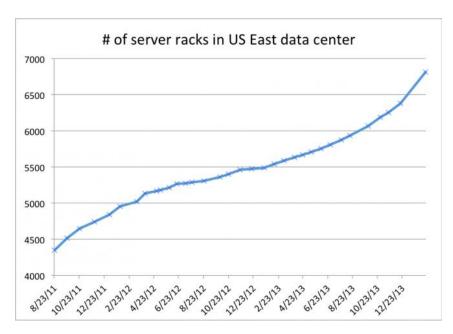


- Google ~ 1 Mil. (2013)
- Microsoft ~ 1 Mil. (2013)
- Facebook ~ 180K (2012)
- OVH ~ 150K (2013)
- Akamai Tech. ~ 127K (2013)
- Rackspace ~ 94K (2013)
- 1&1 Internet ~ 70K (2010)
- eBay ~ 54K (2013)
- HP/EDS ~ 380K (2013)

Source: http://www.datacenterknowledge.com



Facebook Servers

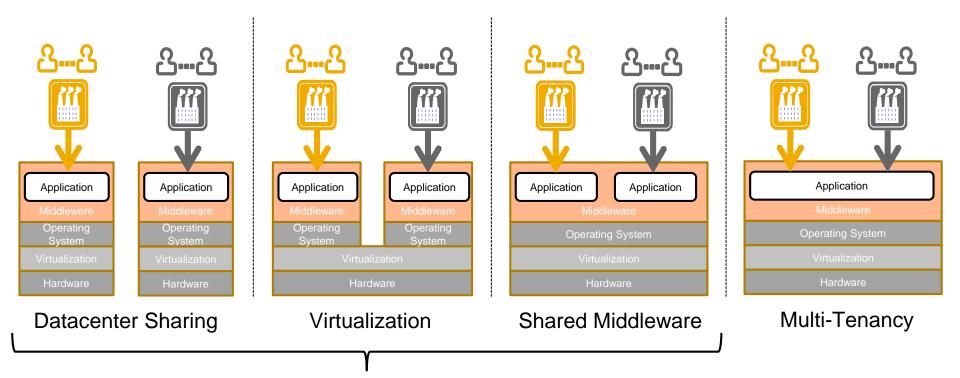


Amazon's Virginia region [Src: Wired.com]



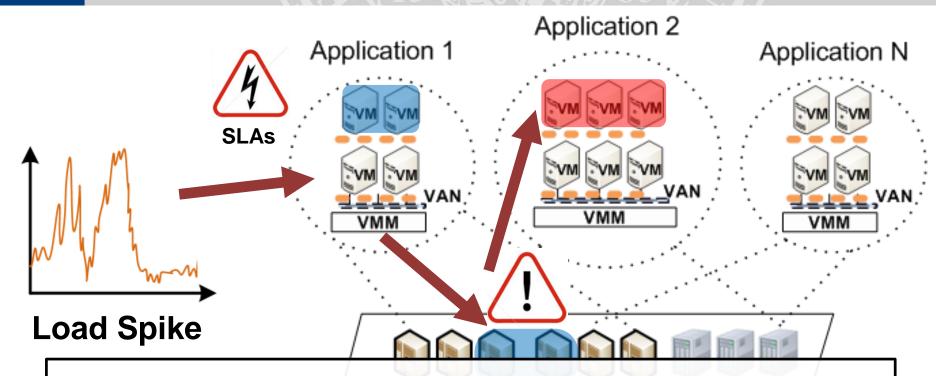
Increasing Pressure to Raise Efficiency

- Proliferation of shared execution environments
- Different forms of resource sharing (hardware and software)
 - Network, storage, and computing infrastructure
 - Software stacks



Multi-Instance

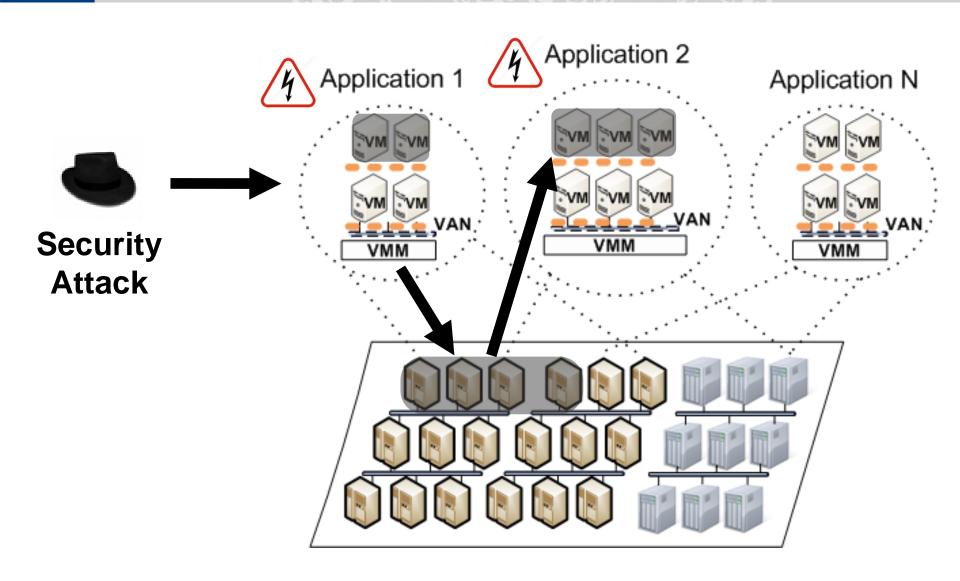




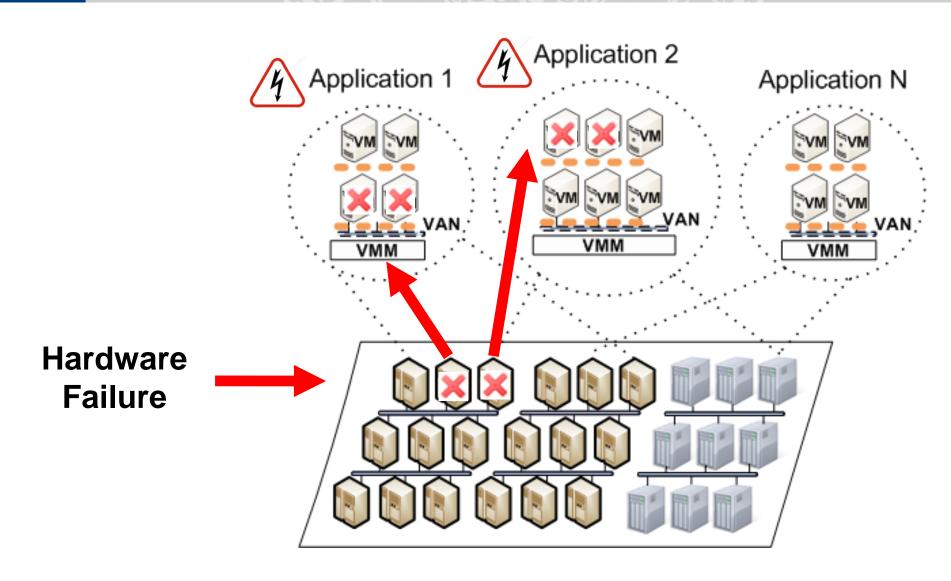
Expand / shrink resources on-the-fly

- When exactly should a reconfiguration be triggered?
- Which particular resources should be scaled?
- How quickly and at what granularity?

Challenges



Challenges





Consequences

- Increased system complexity and dynamics
- Diverse vulnerabilities due to resource sharing
- Inability of to provide dependability guarantees
 - Availability, reliability (+ security, performance, ...)
 - **⇒** Major distinguishing factor between service offerings
- Lack of reliable benchmarks and metrics

"You can't control what you can't measure?" (DeMarco)

"If you cannot measure it, you cannot **improve** it" (Lord Kelvin)



Will What is Needed?

Reliable Metrics

What exactly should be measured and computed?

Representative Workloads

For which scenarios and under which conditions?

Sound Measurement Methodology

How should measurements be conducted?

"To measure is to know." -- Clerk Maxwell, 1831-1879

"It is much easier to make **measurements** than to **know** exactly what you are measuring." -- J.W.N.Sullivan (1928)



The Focus of this Talk

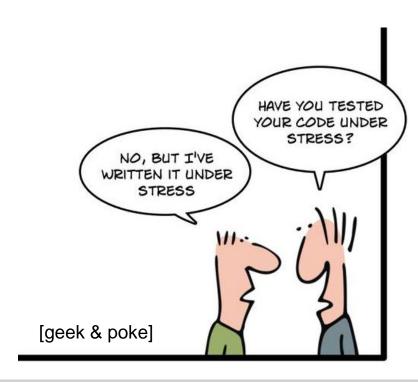
Metrics and benchmarks for quantitative evaluation of

Performance Isolation

- 1. Resource elasticity
- 2. Performance isolation
- 3. Intrusion detection (and prevention)

in shared execution environments

- Virtualized infrastructures
- Multi-tenant applications

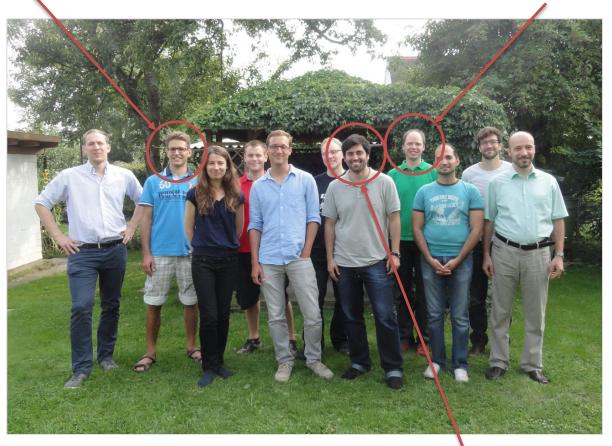


wu Credits

Nikolas Herbst + MSc students

(elasticity)

Rouven Krebs + MSc students (performance isolation)



Aleksandar Milenkoski (intrusion detection)



WILL Part I: Resource Elasticity

Main references

N. Herbst, A. Weber, H. Groenda and S. Kounev. BUNGEE: Benchmarking Resource Elasticity of **Cloud Environments**. Submitted to 6th ACM/SPEC Intl. Conf. on Performance Engineering (ICPE 2015).

N. Herbst, S. Kounev and R. Reussner. Elasticity in Cloud Computing: What it is, and What it is Not. In Proc. of the 10th Intl. Conf. on Autonomic Computing (ICAC 2013), San Jose, CA, June 24-28, 2013. USENIX. [slides | http | .pdf]

Further references

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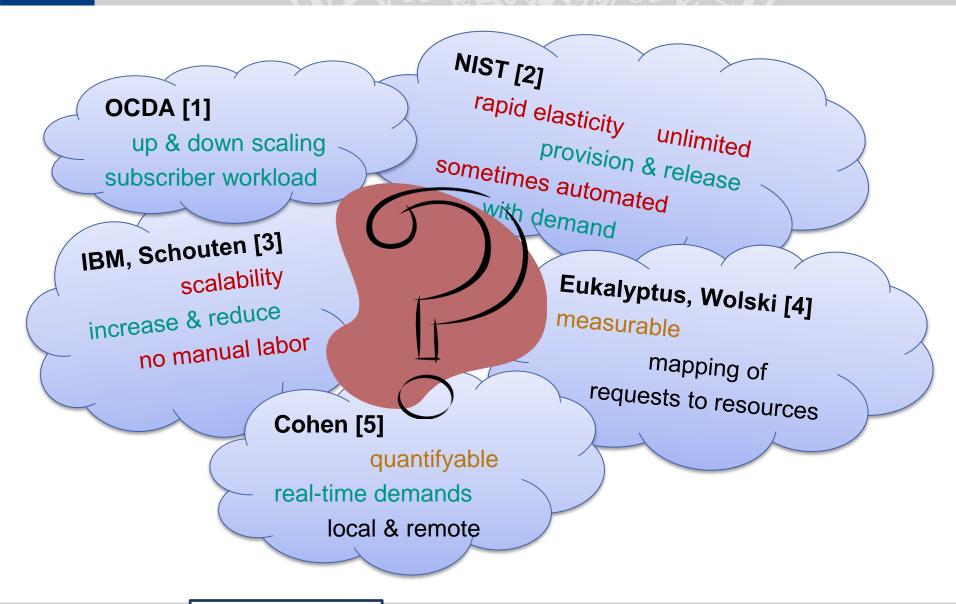
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A. Weber, N. Herbst, H. Groenda and S. Kounev. Towards a Resource Elasticity Benchmark for Cloud Environments. In Proc. of the 2nd Intl. Workshop on Hot Topics in Cloud Service Scalability (HotTopiCS 2014), co-located with ICPE 2014, March 22, 2014. ACM. [slides | .pdf]

Performance Isolation

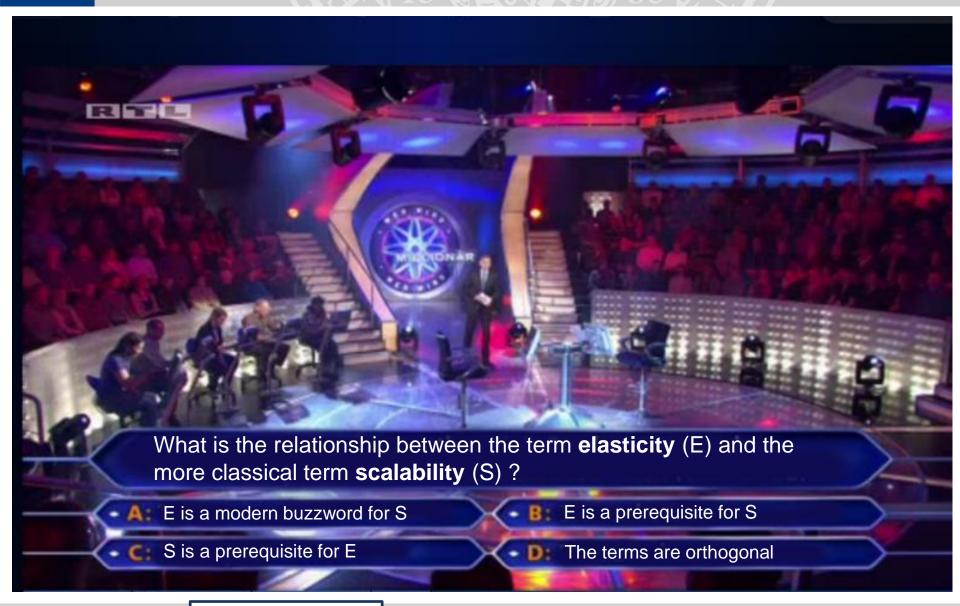


Will What People Say Elasticity is...



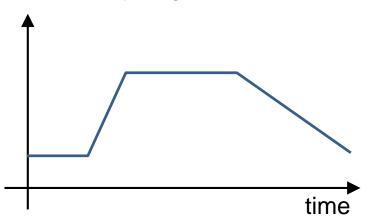


Elasticity vs. Scalability





Workload intensity (e.g., # requests / sec)



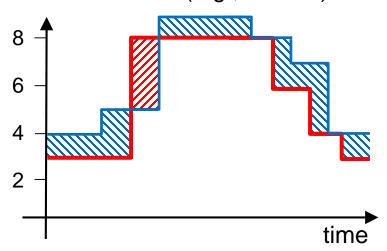
Service Level Objective (SLO)

(e.g., resp. time \leq 2 sec, 95%)

Resource Demand

Minimal amount of resources required to ensure SLOs.

Amount of resources (e.g., # VMs)



resource demand
underprovisioning
resource supply
overprovisioning



Def: The degree to which a system is able to **adapt** to workload changes by provisioning and deprovisioning resources in an autonomic manner, such that at each point in time the available resources match the current demand as closely as possible.

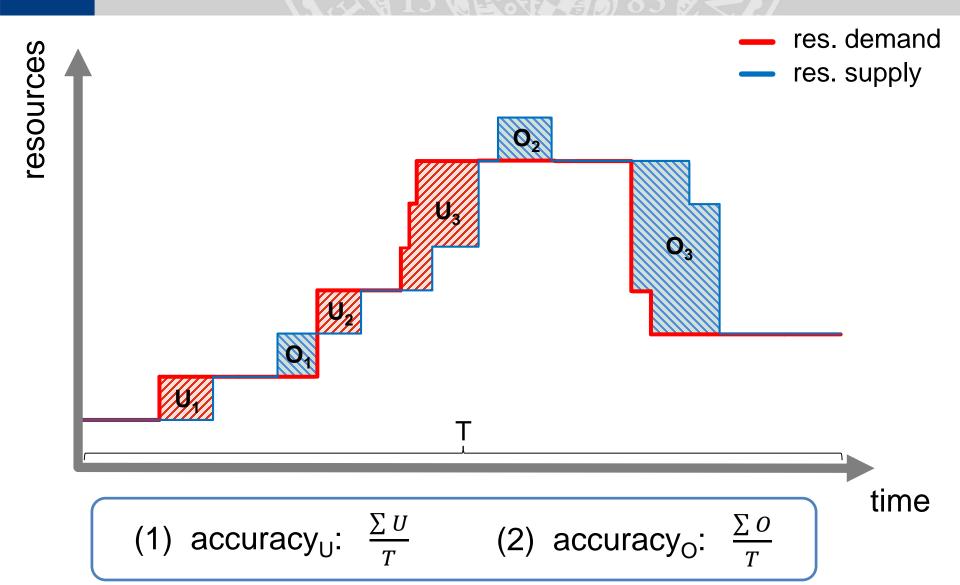
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Performance Isolation

http://en.wikipedia.org/wiki/Elasticity_(cloud_computing)



Metrics: Accuracy

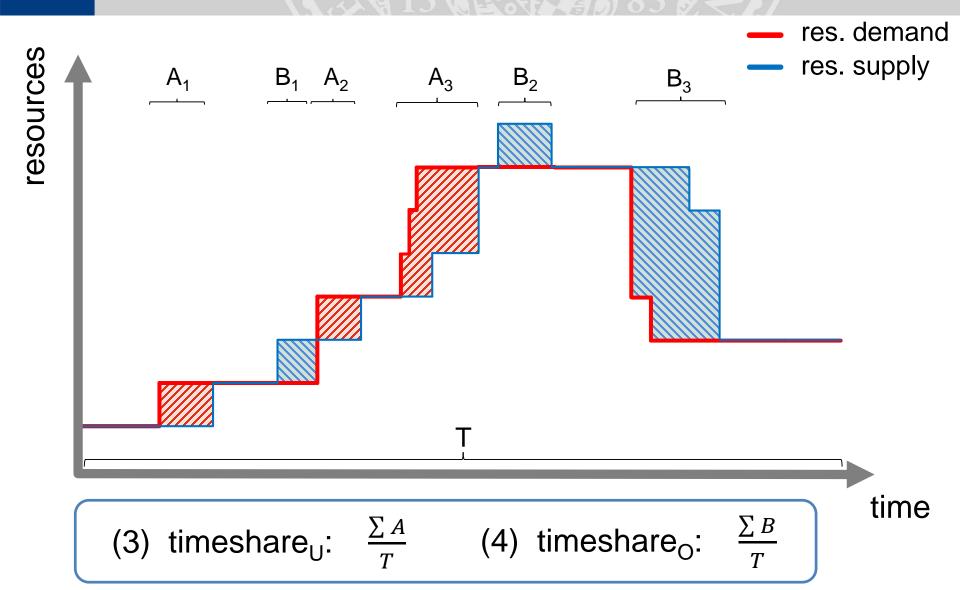


Resource Elasticity

Performance Isolation



Metrics: Timeshare



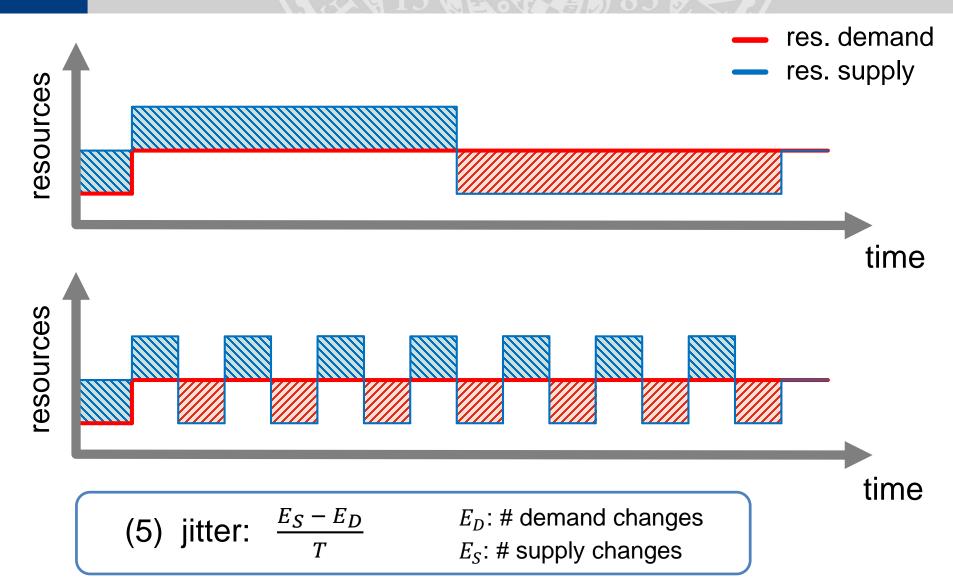
Resource Elasticity

Performance Isolation

Intrusion Detection

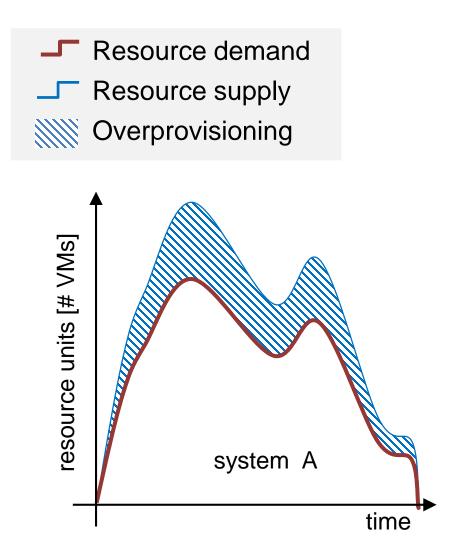


Metrics: Jitter

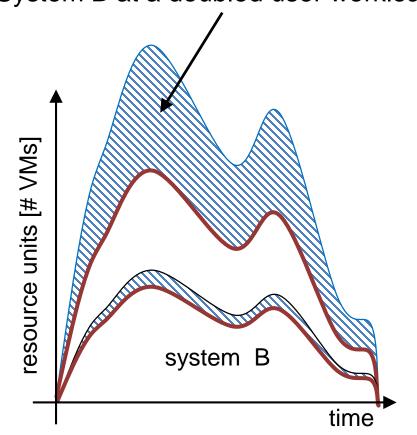




Elasticity Benchmarking

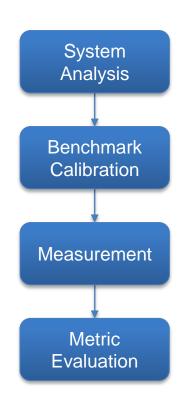


Same user workload on system B System B at a doubled user workload





WU Elasticity Benchmarking Approach



Analyze efficiency & scaling behavior of underlying resources



Adjust load profile



Expose SUT to varying load monitor resource supply & demand



Compute elasticity metrics (accuracy & timing)

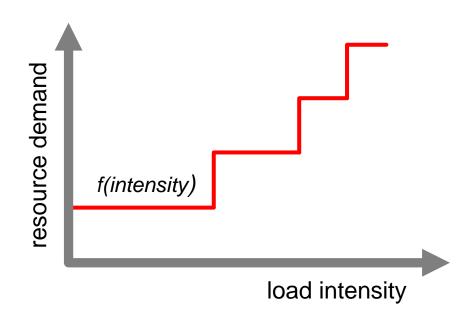


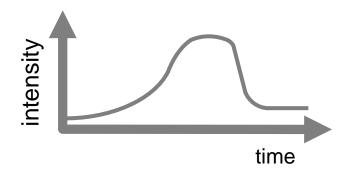
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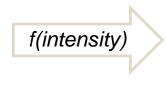


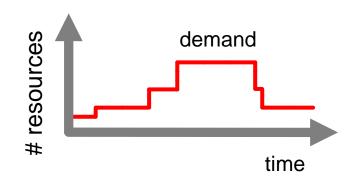
Step 1: System Analysis

- Evaluate system separately at each scale
- Find maximal intensity that the system can withstand without violating SLO (binary search)
- Derive demand step function: resourceDemand = f(intensity)





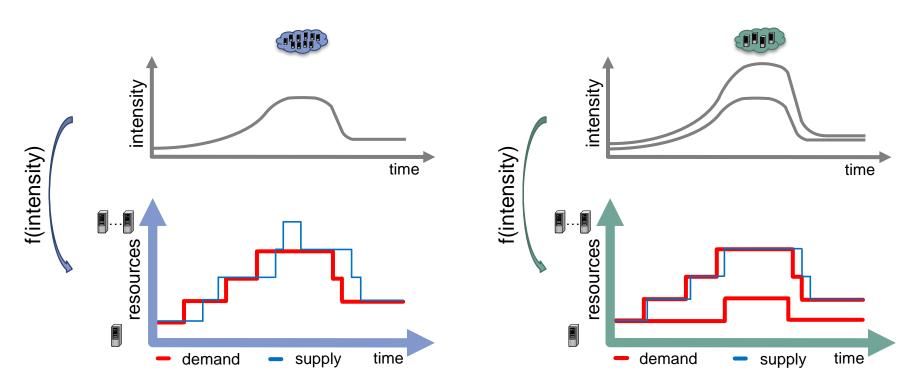






Step 2: Benchmark Calibration

Goal: Induce same resource demand on all systems



- Approach: Adjust load intensity profile to overcome
 - Different efficiency of underlying resources
 - Different scalability



Step 3: Measurement

- Requirement: Stress SUT in a representative manner
 - Realistic variability of load intensity
 - Adaptability of load profiles to suit different domains
- Approach:
 - Open workload model
 - Model load variations with the LIMBO toolkit Facilitates creation of new load profiles
 - Derived from existing traces
 - With desired properties (e.g. seasonal pattern, bursts)
 - Execute load profile using JMeter

Timer-Plugin delays requests according to timestamp file created by LIMBO



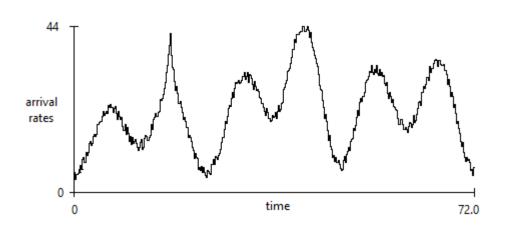




https://github.com/andreaswe/JMeterTimestampTimer



LIMBO: A Tool For Modeling Variable Load Intensities





http://www.descartes-research.net/tools/

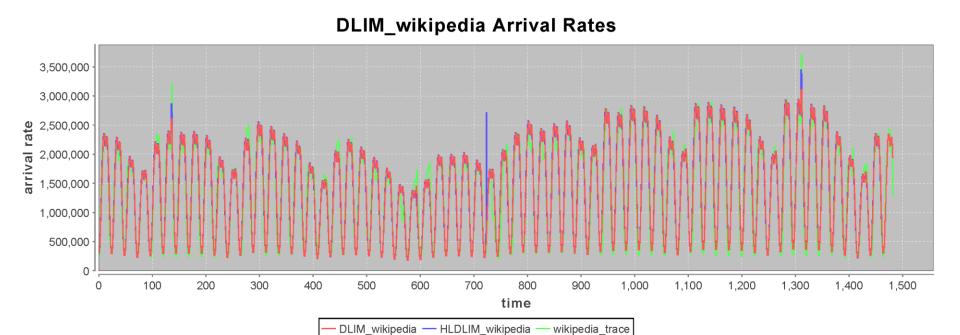
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Slide Template 9/10/2014 27



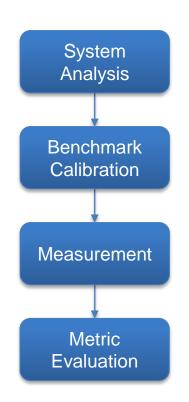
Example: Wikipedia Workload



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WU Elasticity Benchmarking Approach



Analyze efficiency & scaling behavior of underlying resources



Adjust load profile



Expose SUT to varying load monitor resource supply & demand



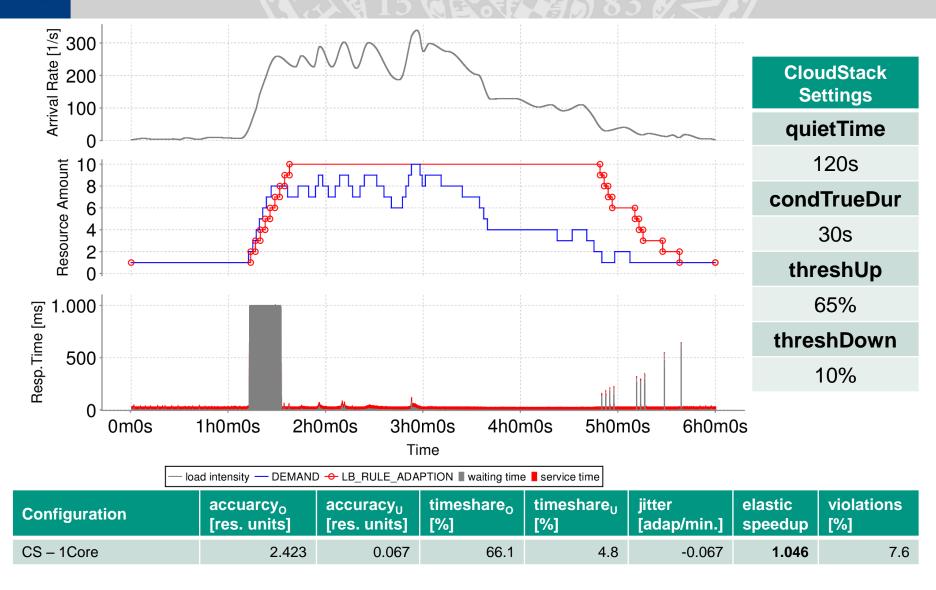
Compute elasticity metrics (accuracy & timing)



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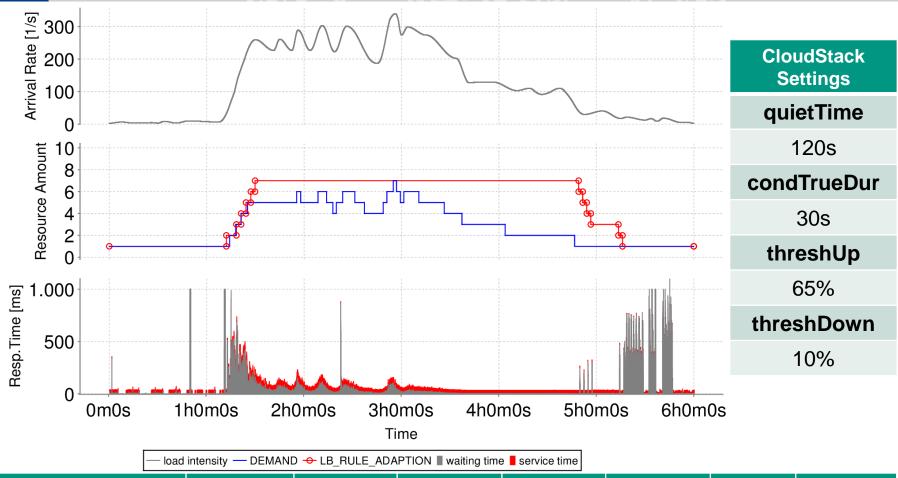
Case Study: CloudStack (CS) - 1Core



Resource Elasticity



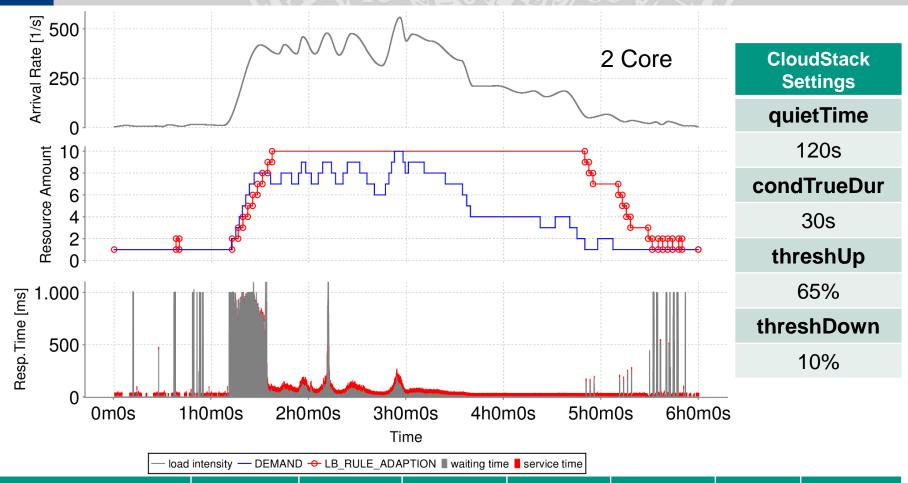
CloudStack (CS) - 2 Core - no adjustment



Configuration	accuarcy _o [res. units]		timeshare _o [%]	timeshare _u [%]	jitter [adap/min.]	elastic speedup	violations [%]
CS – 1Core	2.423	0.067	66.1	4.8	-0.067	1.046	7.6
CS – 2Core no adjustment	1.811	0.001	63.8	0.1	-0.033	1.291	2.1



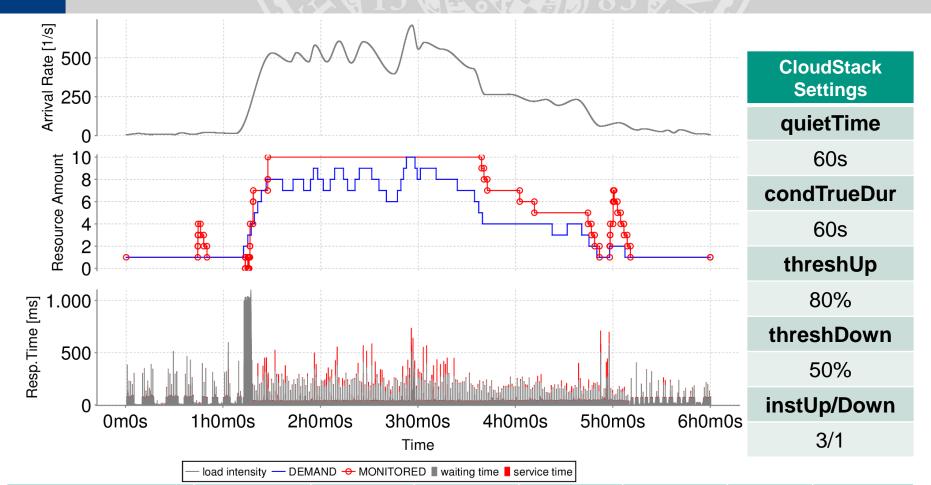
WÜ CloudStack (CS) – 2 Core – adjusted



Configuration	accuarcy _o [res. units]	accuracy _u [res. units]	timeshare _o [%]	timeshare _U [%]	jitter [adap/min.]	elastic speedup	violations [%]
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CS – 2Core adjusted	2.508	0.061	67.1	4.5	-0.044	1.025	8.2



WU Amazon Web Services (AWS) - m1.small



Configuration	accuarcy _o [res. units]	accuracy _u [res. units]	timeshare _o [%]	timeshare _U [%]	jitter [adap/min.]	elastic speedup	violations [%]
CS – 1Core	2.423	0.067	66.1	4.8	-0.067	1.046	7.6
CS – 2Core adjusted	2.508	0.061	67.1	4.5	-0.044	1.025	8.2
AWS - m1.small	1.340	0.019	61.6	1.4	0.000	1.502	2.5



Part II: Performance Isolation

Main references

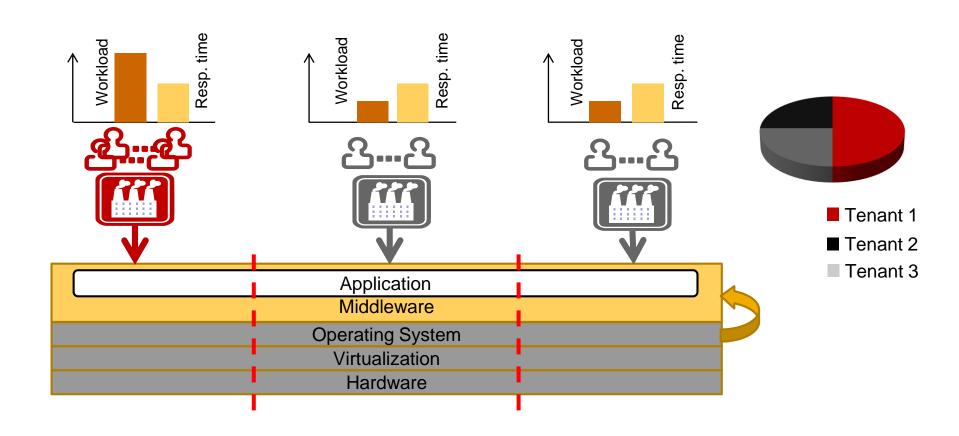
- R. Krebs, C. Momm and S. Kounev. **Metrics and Techniques for Quantifying Performance Isolation in Cloud Environments**. *Elsevier Science of Computer Programming Journal (SciCo)*, Vol. 90, Part B:116-134, 2014, Elsevier B.V. [bib | <u>.pdf</u>]
- R. Krebs, A. Wert and S. Kounev. **Multi-Tenancy Performance Benchmark for Web Application Platforms**. In *Proc. of the 13th Intl. Conf. on Web Engineering (ICWE 2013)*, Aalborg, Denmark, July 8-12, 2013. Springer-Verlag. [.pdf]
- R. Krebs, C. Momm and S. Kounev. **Metrics and Techniques for Quantifying Performance Isolation in Cloud Environments**. In *Proc. of the 8th ACM SIGSOFT Intl. Conf. on the Quality of Software Architectures (QoSA 2012),* Bertinoro, Italy, June 25-28, 2012. ACM. [<a href="http://nxtb.com/http://nxtb.com

Further references

- R. Krebs, S. Spinner, N. Ahmed and S. Kounev. **Resource Usage Control In Multi-Tenant Applications**. In *Proc. of the 14th IEEE/ACM Intl. Symp. on Cluster, Cloud and Grid Computing (CCGrid 2014)*, Chicago, IL, USA, May 26, 2014. IEEE/ACM. [.pdf]
- R. Krebs, M. Loesch and S. Kounev. **Platform-as-a-Service Architecture for Performance Isolated Multi-Tenant Applications**. In *Proc. of the 7th IEEE Intl. Conf. on Cloud Computing*, Anchorage, USA, July 2, 2014. IEEE.
- R. Krebs, C. Momm and S. Kounev. **Architectural Concerns in Multi-Tenant SaaS Applications**. In *Proc. of 2nd Intl. Conf. on Cloud Computing and Services Science (CLOSER 2012)*, Setubal, Portugal, April 18-21, 2012. [<u>.pdf</u>]



Example Scenario: Multi-Tenant Environments

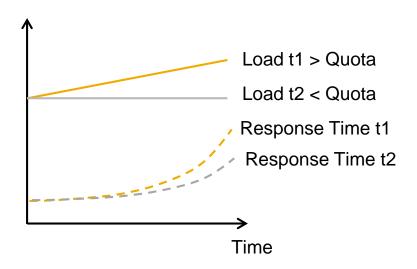


Tenants working within their assigned quota (e.g., # users) should not suffer from tenants exceeding their quotas.



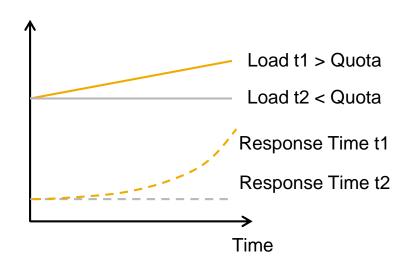
Definition of Performance Isolation

 Tenants working within their assigned quota (e.g., # users) should not suffer from tenants exceeding their quotas.



Non-Isolated System

Resource Elasticity



Isolated System



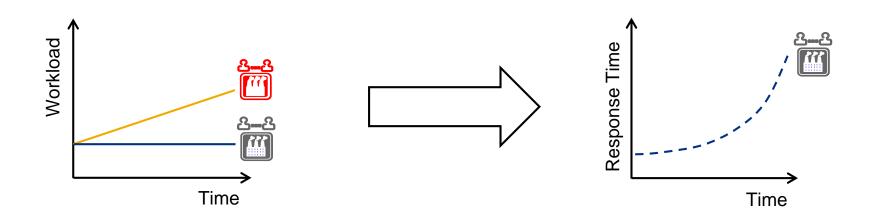
Performance Isolation Metrics



D is a set of disruptive tenants exceeding their quotas.



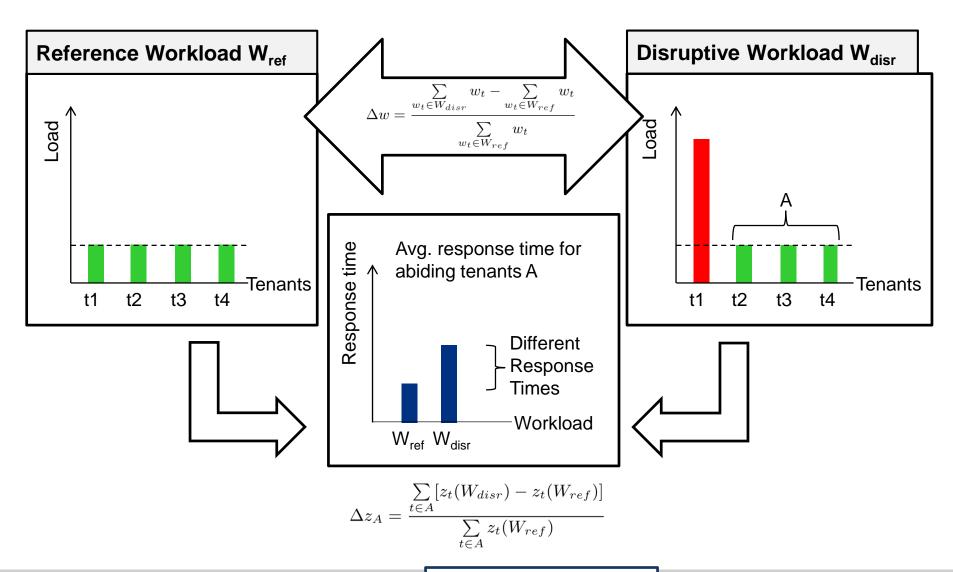
A is a set of **abiding tenants** not exceeding their quotas.



Approach: Quantify impact of increasing workload of the disruptive tenants on the performance of the abiding ones.

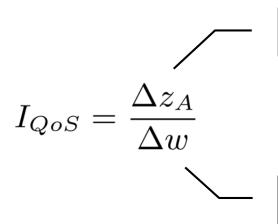


Metrics Based on QoS Impact





WU Example Metric



Difference in response time

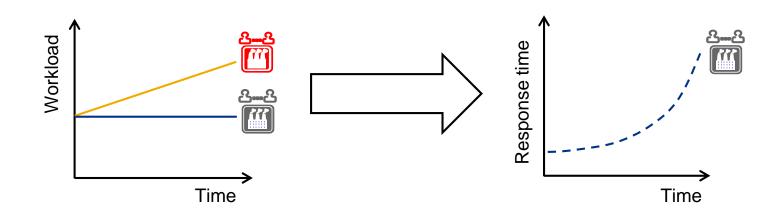
Difference in workload

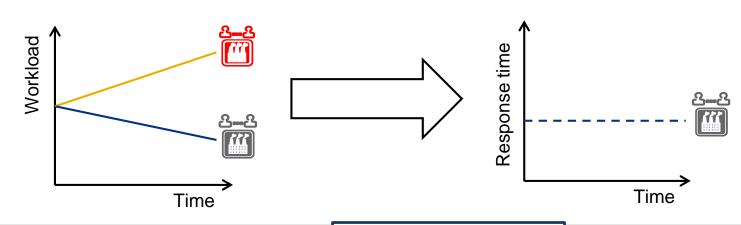
Perfectly Isolated = 0

Non-Isolated = ?

Answers: How strong is a tenant's influence on the others?







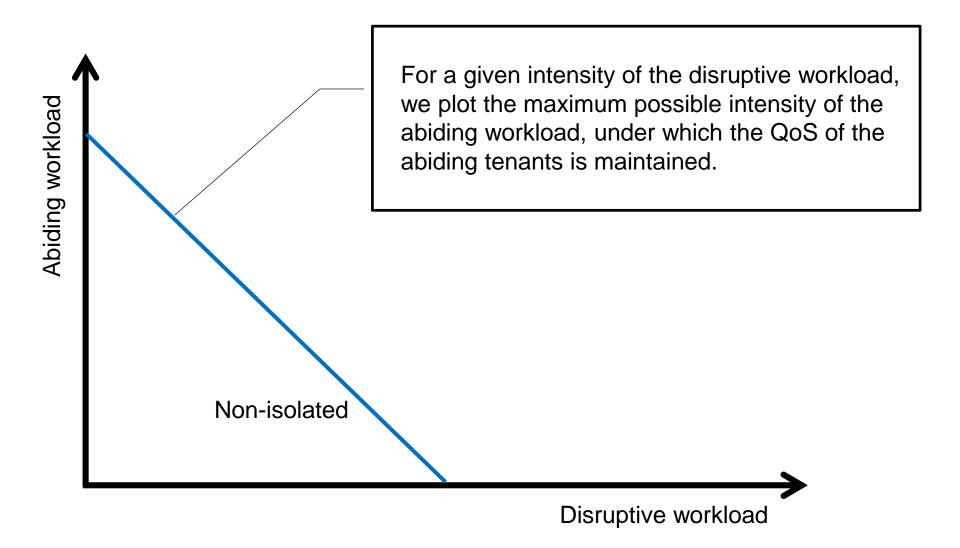
) S. Kounev

Resource Elasticity

Performance Isolation

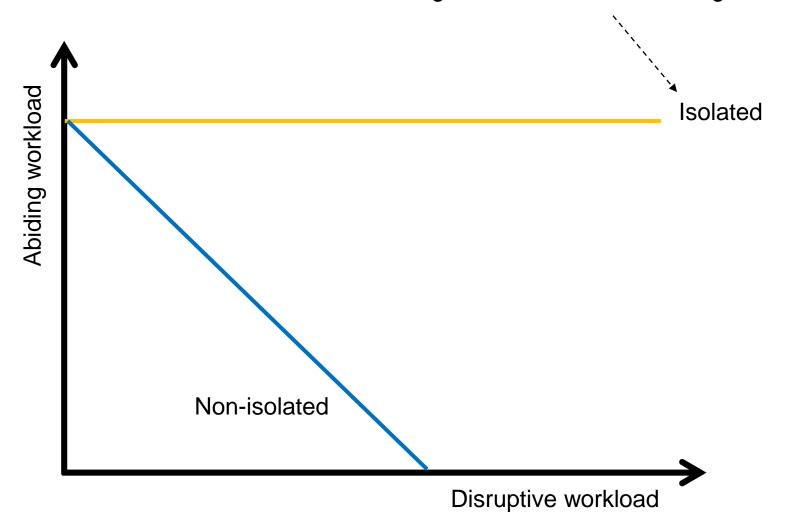
Intrusion Detection





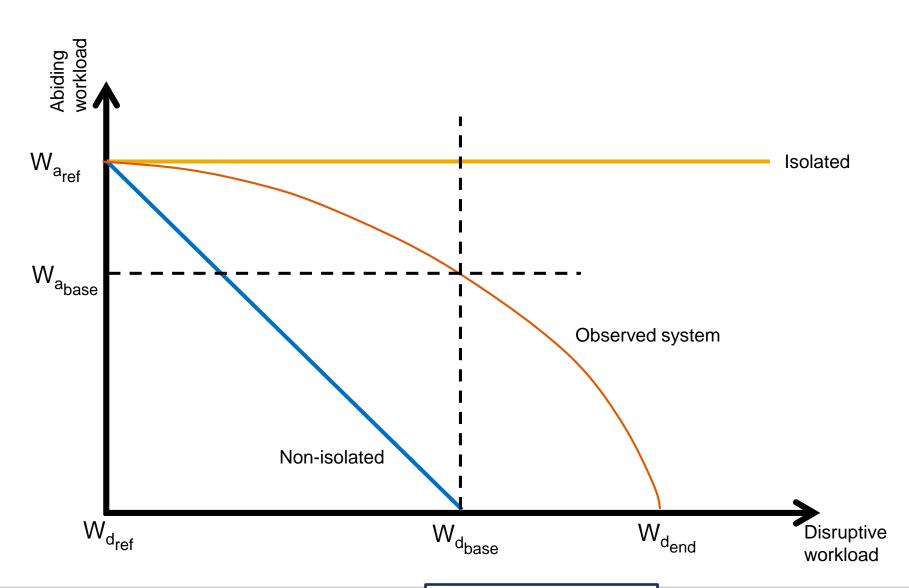


We can maintain the QoS for the abiding tenant without decreasing his workload.



Performance Isolation





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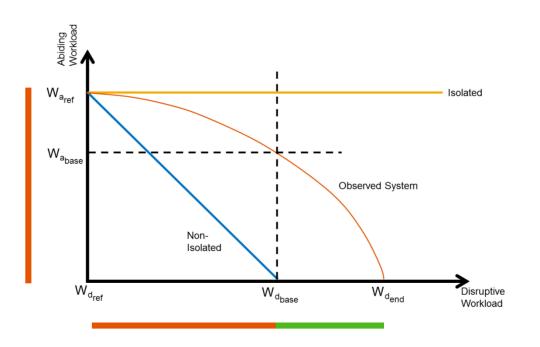
Resource Elasticity

Performance Isolation

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WÜ Example Metric: Iend



$$I_{end} = \frac{W_{d_{end}} - W_{d_{base}}}{W_{a_{ref}}}$$

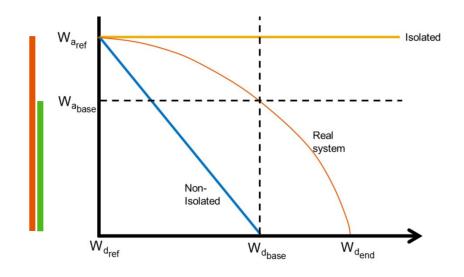
Perfectly Isolated = ?

Non-Isolated = 0

Answers: How isolated is the system compared to a non-isolated system?



WÜ Example Metric: Ibase



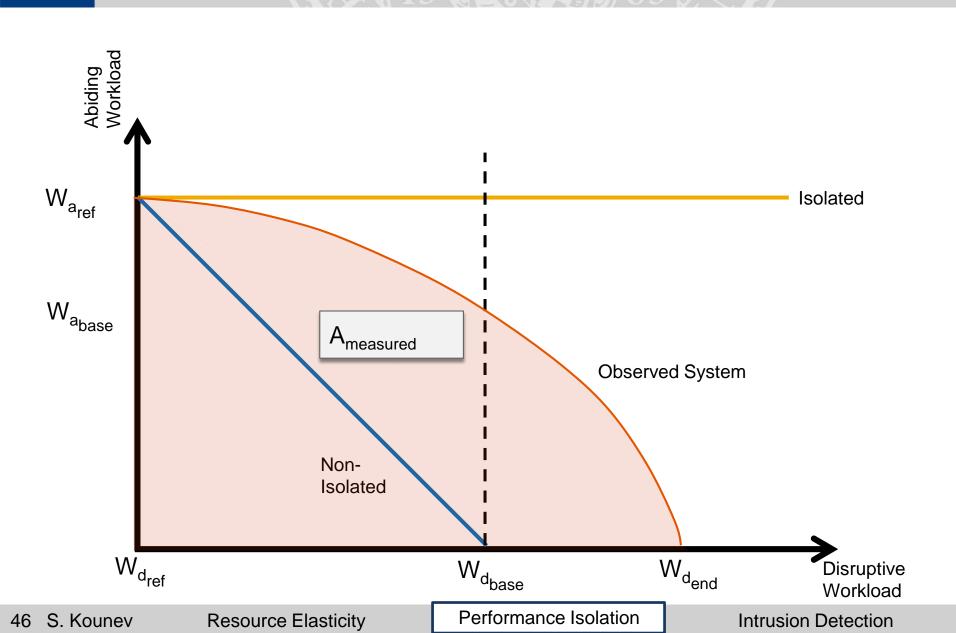
$$I_{base} = \frac{W_{a_{base}}}{W_{a_{ref}}}$$

Perfectly Isolation = 1

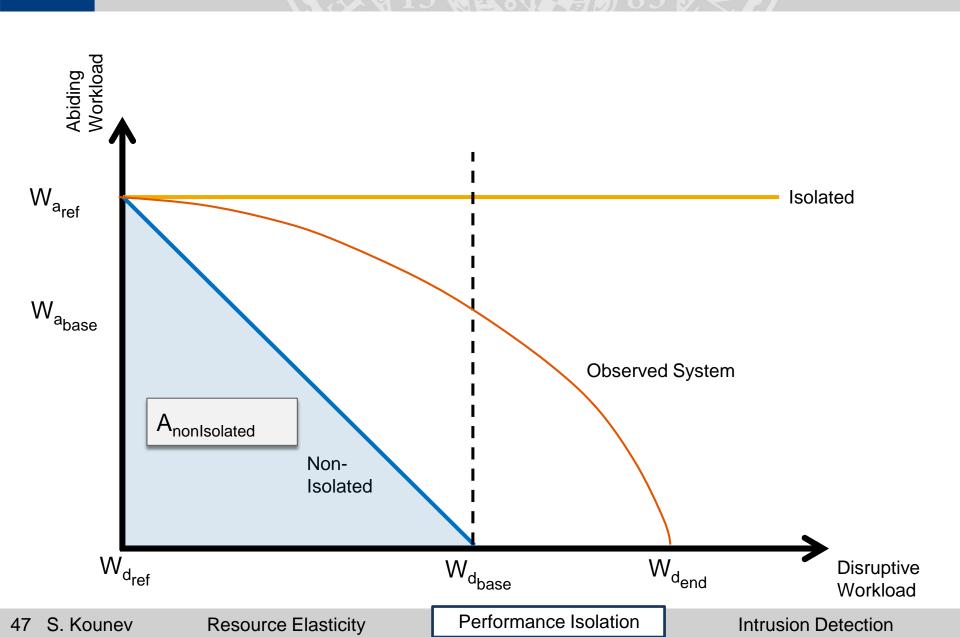
Non-Isolated = 0

Describes the decrease of abiding workload at the point at which a non-isolated systems abiding load is 0.

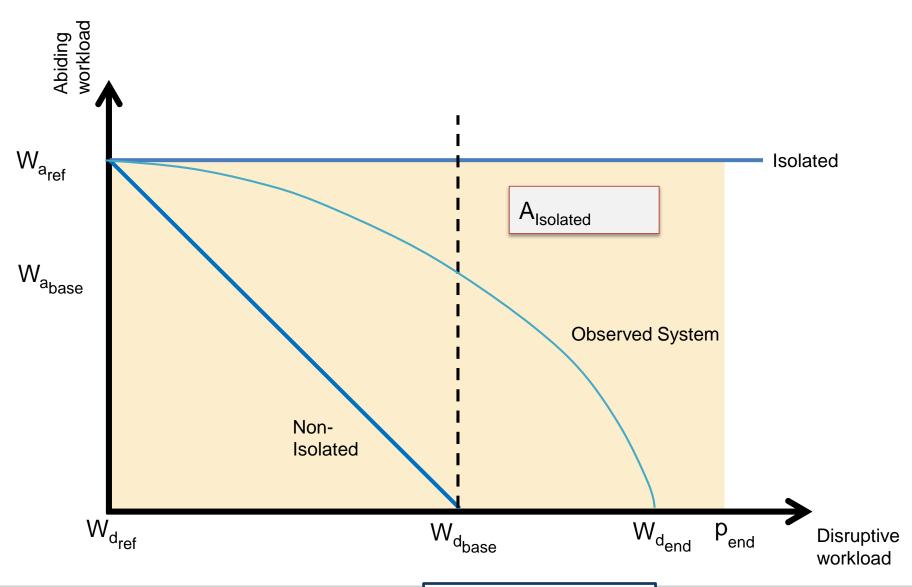












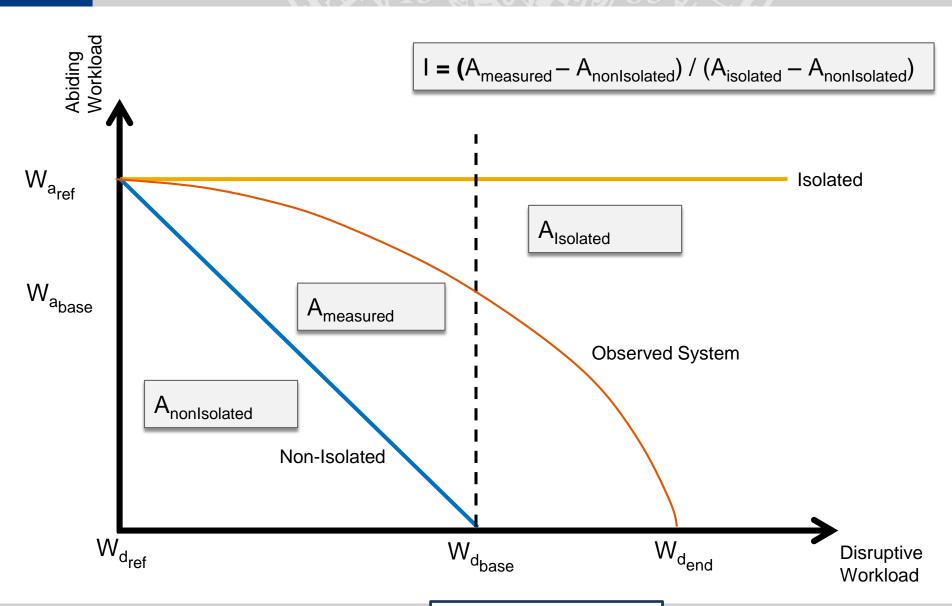
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Resource Elasticity

Performance Isolation

Intrusion Detection





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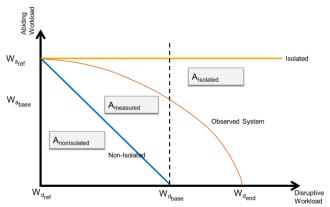
Resource Elasticity

Performance Isolation

Intrusion Detection



WÜ Example Metrics: IntBase and IntFree



$$I_{intBase} = rac{inom{W_{d_{base}}}{\int W_{d_{ref}}} f_m(W_d) dW_d}{W_{d_{ref}}} - W_{a_{ref}}^2/2} \qquad ext{Areas within $W_{ ext{d}_{ref}}$ and $W_{ ext{d}_{base}}$}$$

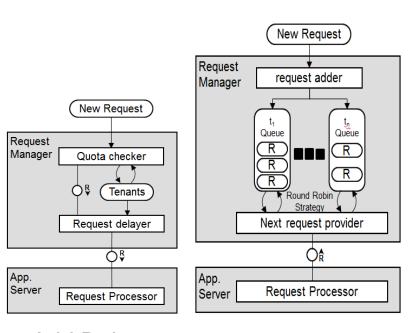
$$I_{intFree} = rac{ \left(\int \limits_{W_{d_{ref}}}^{p_{end}} f_m(W_d) dW_d
ight) - W_{a_{ref}}^2/2}{W_{a_{ref}} \cdot (p_{end} - W_{d_{ref}}) - W_{a_{ref}}^2/2} \qquad ext{Areas within $W_{ ext{d}_{ref}}$ and predefined bound.}$$

Perfectly Isolated = 1

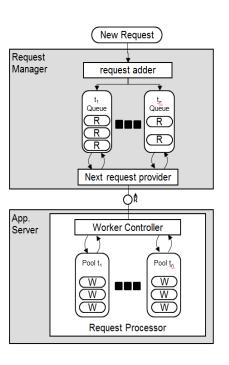
Non-Isolated = 0

Answers: How much potential has the isolation method to improve?





New Request Request Quota checker Tenants Manager request adder FIFO Normal queue always first Next request provider Server Request Processor



Add Delay

Round Robin

Blacklist

Separate Thread Pools

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Part III: Intrusion Detection

Collaboration with

Marco Vieira and Nuno Antunes, University of Coimbra, Portugal Bryan D. Payne, Department of Security Research, Nebula Inc. Alberto Avritzer, Siemens Corporate Research, USA

Main references

A. Milenkoski, B. Payne, N. Antunes, M. Vieira and S. Kounev. **An Analysis of Hypercall Handler Vulnerabilities**. In *Proc. of 25th IEEE Intl. Symp. on Software Reliability Engineering (ISSRE 2014) - Research Track*, Naples, Italy, November 2014. IEEE.

A. Milenkoski, B. Payne, N. Antunes, M. Vieira and S. Kounev. **HInjector: Injecting Hypercall Attacks for Evaluating VMI-based Intrusion Detection Systems** (Poster Paper). In *2013 Annual Computer Security Applications Conf. (ACSAC 2013)*, New Orleans, Louisiana, USA, 2013. [.pdf]

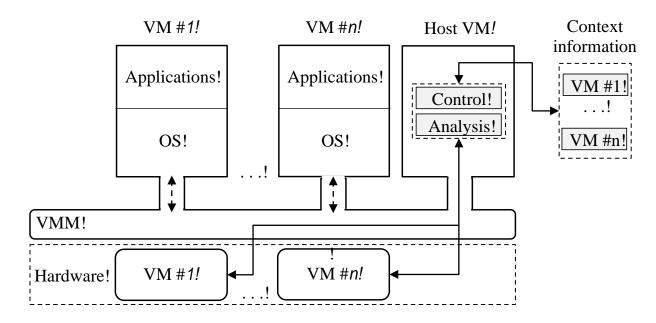
Further references

A. Milenkoski, S. Kounev, A. Avritzer, N. Antunes and M. Vieira. **On Benchmarking Intrusion Detection Systems in Virtualized Environments**. Technical Report SPEC-RG-2013-002 v.1.0, SPEC Research Group - IDS Benchmarking Working Group, Standard Performance Evaluation Corporation (SPEC), June 2013. [<u>.pdf</u>]

A. Milenkoski, M. Vieira, B. Payne, N. Antunes and S. Kounev. **Technical Information on Vulnerabilities of Hypercall Handlers**. Technical Report SPEC-RG-2014-001 v.1.0, SPEC Research Group - IDS Benchmarking Working Group, Standard Performance Evaluation Corporation (SPEC), August 2014. [<u>.pdf</u>]



- Evaluation of intrusion detection systems (IDSes)
 - Enables the comparison of IDSes
 - Enables the improvement of the configuration of a deployed IDS
- IDSes for virtualized environments → many designs possible
 - Network intrusion detection by monitoring the virtual network bridge
 - Host intrusion detection through Virtual Machine Introspection (VMI)



Performance Isolation

53 S. Kounev



WI Focus of our Work

IDS evaluation in virtualized environments

Workloads



Injection of attacks targeting VMMs



Injection of representative hypercall attacks

Metrics and measurement methodologies



New security-related metrics



Attack detection accuracy metrics that take elasticity into account



Malicious Workloads: Generating Attacks

- Focus: VMMs as attack surfaces
 - Attack scenario: "malicious guest VM attacks the underlying VMM"
 - Attack vectors

Hypercalls VM device drivers VM exits

- Hypercalls
 - Routines / software traps invoked by kernels of paravirtualized, or HV with paravirtualized device(s), guest VMs for performing system management operations (e.g., sharing memory pages)

 system call
 hypercall

 User-mode applications
 OS
 Guest VM's OS
 VMM

Vulnerabilities in VMMs' hypercall handling routines are critical!



Malicious Workloads: Generating Attacks

- Defining representative/realistic attack scenarios
 - Attack models
 - Identify characteristics of hypercall attacks (e.g., specific hypercall parameter values, hypercall order,)
 - No attack scripts/proof-of-concept code available ...
 - ... however, patches are available!
- Approach:
 - 1. Select a set of hypercall vulnerabilities
 - 2. Reverse-engineer the patches of the selected vulnerabilities

2.1 Develop proof-of-concept code

3. Characterize hypercall attacks



Malicious Workloads: Generating Attacks

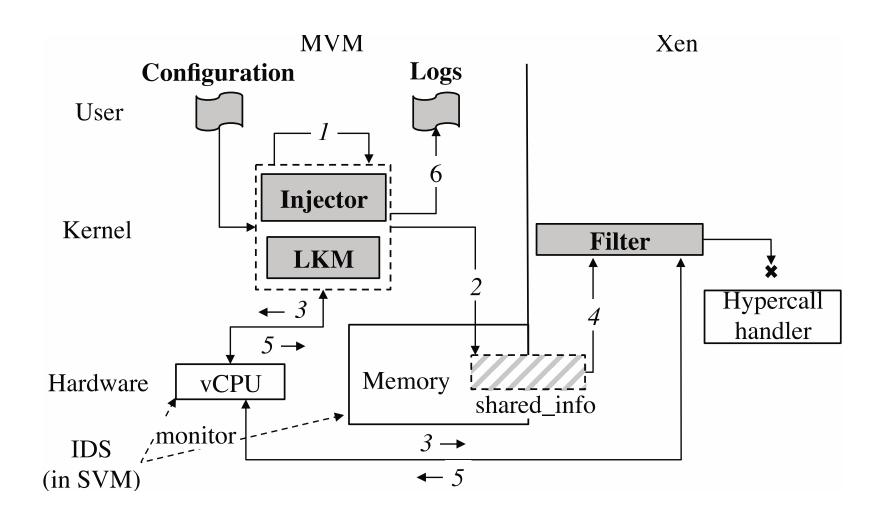
- Artificial injection of hypercall attacks based on representative attack models
 - Reason: Lack of publicly available attack scripts
- Attack models
- Analysis of relevant CVE reports
- 2. Identification of patterns of VM activities
- 3. Categorization of VM activity patterns into attack models

- Attack patterns
- 1. Invoking hypercalls from irregular call sites
- 2. Hypercalls with anomalous parameter values a) outside the valid value domains, or b) crafted for exploiting specific vulnerabilities (not necessarily outside the valid value domains)
- 3. A series of hypercalls in irregular order, including repetitive execution of a single or multiple hypercalls

More later ...



HInjector: Framework for Injecting Hypercall Attacks





Field Study on Hypercall Vulnerabilities

Goals

- Characterization and classification of hypercall vulnerabilities
- Identification of causes of hypercall vulnerabilities
- Provide technical information on hypercall vulnerabilities

Benefits

- Can we prevent future vulnerabilities?
 - Hypercall programming practices
 - Vulnerability discovery techniques
- Can we detect and prevent the exploitation of existing vulnerabilities?
 - Hypercall attack detection and prevention mechanisms



Field Study on Hypercall Vulnerabilities

CVE	Hypercall	Vulnerable Platform
CVE-2012-3497 / CVE-2012-6036	tmem_op	>= Xen 4.0.x
CVE-2012-5513	memory_op	< Xen 4.1.4
CVE-2008-3687	flask_op	< Xen 3.3
CVE-2013-0154	mmu_update	Xen 4.2.x
CVE-2013-1964	grant_table_op	Xen 4.1.x – 4.1.5
CVE-2012-4539	grant_table_op	Xen 4.1.x – 4.1.4
CVE-2012-5525	mmuext_op	Xen 4.2.x
CVE-2012-5515	memory_op	Xen 3.4.x – 4.1.4
CVE-2012-3494	set_debugreg	< Xen 4.1.4 (4.1 ser.), Xen 4.2.0 (4.2 ser.)
CVE-2012-3496	memory_op	Xen 3.9.x – 4.1.4
CVE-2012-5514	memory_op	Xen 3.4.x – 4.1.4
CVE-2012-3495	physdev_op	Xen 4.1.x
CVE-2013-0154	mmuext_op	Xen 4.2.x
CVE-2012-5513	memory_op	Xen 4.1.x
CVE-2013-4553	domctl	> Xen 3.4.x
CVE-2013-0151	hvm_op	Xen 4.2.x
CVE-2013-4494	grant_table_op	All versions of Xen up to the current date
CVE-2012-5510	grant_table_op	< Xen 4.1.4 (4.1 ser.), Xen 4.2.0 (4.2 ser.)
CVE-2013-3898	unknown	Windows 8 / Windows Server 2012

Intrusion Detection



- Errors causing hypercall vulnerabilities
 - Implementation errors (missing value validation, incorrect value validation, and incorrect implementation of inverse procedures)
 - Hypervisor design errors
- Most implementation errors are missing value validation errors
 - Internal variables (e.g., return codes) !
 - Eliminating missing value validation errors by adding program code verifying variable values ->
 - Reduces hypercall execution speed → increased frequency of continuations → performance overhead →
 - Programming practices for boosting hypercall execution speed → vulnerabilities (e.g., CVE-2012-5535)



Field study on hypercall vulnerabilities: Observations (cont.)

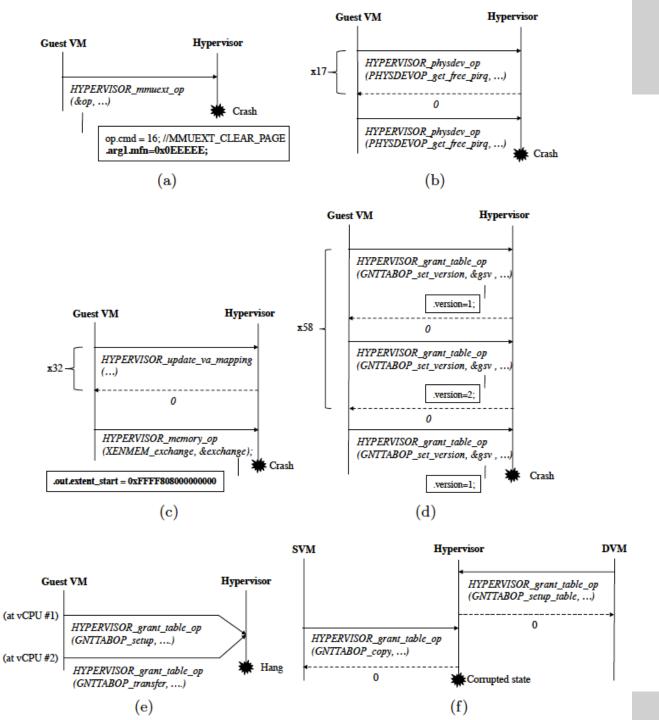
Hypercall attacks

- Effects: crash, hang, corrupt state, information leakage
- Very effective hypervisor DoS attacks critical: downtime minute of the virtualized cloud infrastructure of Amazon costs \$66,240
- An effective mechanism for intruding hypervisors, however, as part of a multi-step attack
 - Hypercall attack -> paving the way for further malicious activities

Hypercall attack models

- execution of a single hypercall with:
 - regular parameter value(s) (i.e., regular hypercall), or
 - parameter value(s) specifically crafted for triggering a given vulnerability, which includes values inside and outside valid value domains, or
- execution of a series of regular hypercalls in a given order, including:
 - repetitive execution of a single hypercall, or
 - repetitive execution of multiple hypercalls.
- where an execution of (a) regular hypercall(s) is performed in a way such that:
 - the targeted hypervisor cannot properly handle by design, or
 - an erroneous program code is reached.







Standard-Performance-Evaluation-Corporation

Open-Systems-Group (OSG)

- Processor and computer architectures
- Virtualization platforms
- Java (JVM, Java EE)
- Message-based systems
- Storage systems (SFS)
- Web-, email- and file server
- SIP server (VoIP)
- Cloud computing

High-Performance-Group (HPG)

- Symmetric multiprocessor systems
- Workstation clusters
- Parallel and distributed systems
- Vector (parallel) supercomputers

"Graphics and Workstation Performance Group" (GWPG)

- CAD/CAM, visualization
- OpenGL







WU SPEC Research Group (RG)

- Founded in March 2011
 - Transfer of knowledge btw. academia and industry
- **Activities**
 - Methods and techniques for experimental system analysis
 - Standard metrics and measurement methodologies
 - Benchmarking and certification
 - Evaluation of academic research results
- Member organizations (Feb 2014)



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University of Minnesota

Driven to Discover

IIT Bombay





Christian-Albrechts-Universität zu Kie









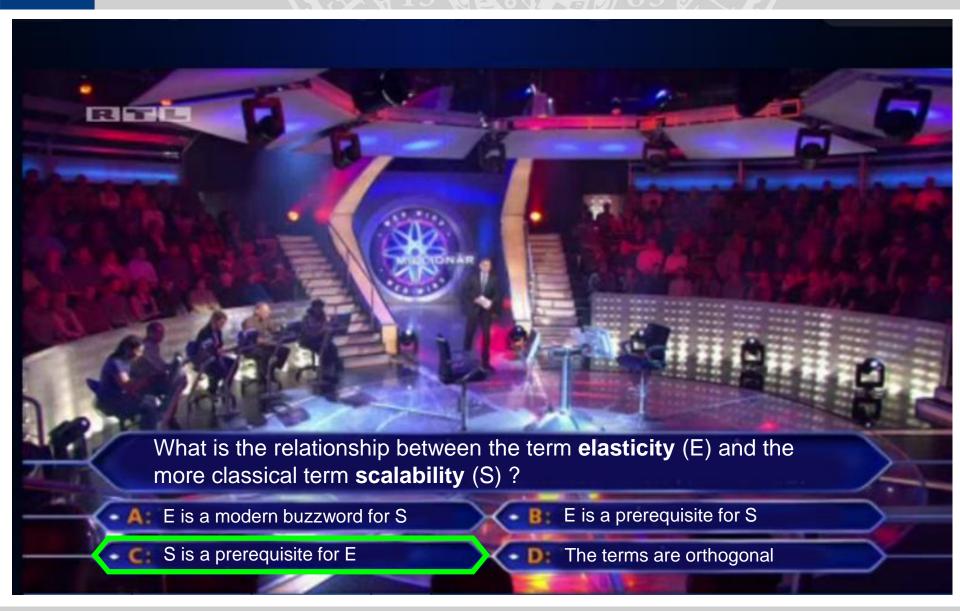








Elasticity vs. Scalability





Thank You!

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