

#### Catching Falling Dominoes: Cloud Management-Level Provenance Analysis with Application to OpenStack

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### Outline

- Cloud Security Challenge
- Limitation of Exiting Solutions
- Cloud Provenance Model
- Methodology
- Implementation
- Experiment Results
- Conclusion

 Cloud computing has been widely adopted to provide users the ability to selfprovision resources while optimally sharing the underlying physical infrastructure



Tenants

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- The self-service and multi-tenancy nature of clouds also leads to a higher complexity and greater chances of misconfigurations
- Adversarial actors can launch their attack by exploiting cloud misconfigurations
- This makes explaining systems behaviour difficult → The way to find the root cause

 The self-service and multi-tenancy nature of clouds also leads to a higher complexity and greater chances of misconfigurations



### The Need for Root Cause Analysis

- The need for finding the root cause:
  - Forensic analysis
  - Debugging
  - Prevention of recurrent failures
  - Recovery

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## Limitation of Existing Solutions

#### Problem localization solutions

- E.g., using alert correlation [1]
- Not providing root cause operations

#### Cloud logs investigation

- No intrinsic central view of changes in different services
- Log aggregation cannot explain the interdependencies between events
- Logs are not sufficiently expressive

### Limitation of Existing Solutions

#### Existing provenance solutions

- On Low-level system calls [2, 3] and not sufficient for clouds
  - $\odot \text{Big}$  size of generated records
  - ${\scriptstyle \odot}\mbox{Tedious}$  and error-prone analyses
  - Storage and network overhead
- Impractical interception mechanisms in clouds
   Requiring system instrumentation



### **Example Attack Scenario**

- A data leakage alert is released
- Based on the logs, the cloud admin cannot explain the complex chain of events leading to the attack
  - Interdependency
  - Expressiveness



### **Example Attack Scenario**

#### • What if we could find the interdependencies?



### DOMINOCATCHER

- The first provenance model at cloud management level
- A mechanism to capture the provenance metadata from different services in clouds and construct the provenance graph
  - A less invasive interception mechanism
  - Incremental construction
- Provenance-based forensic approach
  - User preference-based pruning
- Implementation based on OpenStack

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### **Cloud Management Provenance Model**

- Defined based on a standard provenance specification [4] where:
  - Nodes:
    - Entities: virtual resources
    - Activities: cloud management operations
    - Agents: cloud tenants or users



### **Cloud Management Provenance Model**

- Defined based on a standard provenance specification [4] where:
  - Edges: The provenance edges encode the dependencies between nodes. E.g.,
    - Used, WasGeneratedBy (both pointing to the past in time)



### **Cloud Management Provenance Model**

- VMA's network is attached to TenantA's router
- A TenantB's user created a port on that router
  - So, that user could enter TenantA's network
- The port was updated once a new VM was attached to it
  - So, that user could disable anti-spoofing rules to launch the attack
    WasGeneratedBy
    Create
    WasGeneratedBy



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### DOMINOCATCHER Methodology



### **Provenance Construction**



## Provenance Construction – Data Collection



# Provenance Construction – Data Collection



# Provenance Construction – Data Collection



# Provenance Construction – Graph Generation



# Provenance Construction – Graph Generation



### **Forensic Analysis**



### Forensic Analysis – Challenges

- Large size of the provenance graph
- Different tenants' analysis requirements



Forensic analysis only on this tenant users' behaviour



Forensic analysis on all tenants with whom there was a traffic exchange



- 1. Disjoint subgraph
- 2. Context-based

- 1. Disjoint subgraph
  - Discards the subgraphs to which there is no path from the target resource node

1. Disjoint subgraph



- 2. Context-base
  - Traverses paths while checking the specified constraints to identify a subgraph of resources and operations interdependent with the victim virtual resource

2. Context-base



### **Forensic Analysis**



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### Implementation

- Implemented on OpenStack (Rocky)
- Neo4j as the graph database and Cypher language to query
- Deployed as WSGI middlewares on OpenStack services
  - Configuration changes are performed by these services → Capturing all configuration changes through API calls
  - As an attached interface requires less customization →
     Less invasiveness
- Py2neo library to translate python queries into Cypher

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### **Experiment Results**

- Measured the ratio between the added latency and OpenStack management operations execution time in various cloud sizes
  - Total overhead remains under %4.17

Cloud Size	# of Provenance Graph Nodes	Data Collection	Graph Generation	Total Overhead
600 VMs	43069	%.21	%1.89	%2.10
1800 VMs	64689	%.23	%3.32	%3.56
3000 VMs	107936	%.23	%3.94	%4.17

### **Experiment Results**

- Measured the ratio between the added latency and OpenStack management operations execution time in various cloud sizes
  - Total overhead remains under %4.17

![](_page_38_Figure_3.jpeg)

### **Experiment Results**

- For a provenance graph constructed with 120,000 operations, only 80-megabyte storage is required
  - Higher than the number of configuration API calls issued in one day in a real enterprise cloud

![](_page_39_Figure_3.jpeg)

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### **Concluding Remarks**

- Defined a provenance model on cloud management level
- Provided an interception mechanism deployed in different cloud services
  - Less invasiveness
- Proposed a provenance-based forensic analysis approach for clouds
- Implemented and evaluated in OpenStack

### References

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### **Thanks & Questions**

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